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ARRAY SCANNING FOR IMAGE TRANSMISSION.(U)  
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Final Technical Report  
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# ARRAY SCANNING FOR IMAGE TRANSMISSION

Harris Corporation

Dr. W. C. Adams, Jr.  
Dr. T. R. Bader  
Dr. J. J. Knab  
Dr. E. B. Knick, Jr.  
Dr. C. A. Shuman

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APPROVED:

*Leonard E. Converse, Jr.*

LEONARD E. CONVERSE, Jr.  
Project Engineer

APPROVED:

*Howard Davis*

HOWARD DAVIS  
Technical Director  
Intelligence & Reconnaissance Division

FOR THE COMMANDER:

*John P. Huss*

JOHN P. HUSS  
Acting Chief, Plans Office

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recording feasibility evaluation, are used to formulate the trade-offs necessary to provide a conceptual system design and implementation. The conceptual design that is presented fully exploits the nature of the array scanning. Various practical aspects and system considerations regarding the conceptual design are treated.

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## EVALUATION

The final report on this study effort provides information which can be utilized for the development of a system for transmitting imagery or in evaluating tradeoffs for scanning and recording image information, compressing image data, and restoring and enhancing image data. This forms a basis for meeting the objectives of RADC TPO #R2C, which is concerned with each of the above image related problems.

*Leonard E. Converse, Jr.*  
LEONARD E. CONVERSE, JR.  
Project Engineer

## 1.0

## INTRODUCTION

This final report contains the documentation of the results of the Array Scanning for Image Transmission Study performed in accordance with RADC contract number F30602-77-C-0181 during the contract period September 16, 1977 to September 20, 1978.

In this introduction we provide an overview and description of the nature of the study effort, followed by a description of the study approach we have followed. This is followed by a summary of the organization of the final report. The introduction concludes with a summary of the conclusions of the study.

### 1.1


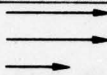

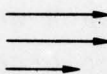
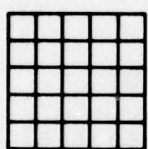
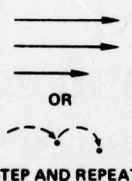
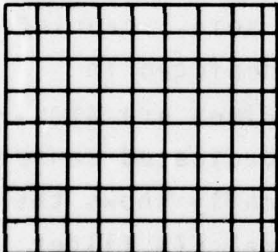
#### AN OVERVIEW OF THE NATURE OF THE STUDY

Figure 1.1-1 shows the hierarchy of spatial sampling or scanning techniques of increasing generality and complexity ranging from the conventional point scanning method of Figure 1.1-1 (a) to the situation depicted in Figure 1.1-1 (d) in which all of the pixel values are simultaneously available, i.e., there exists one dedicated sensor for each image pixel. Figure 1.1-2 schematically shows the implementation of a generalized two-dimensional FIR filter which operates on the output of a two-dimensional sensor array whose pixel output values are all simultaneously available.

The motivation for this study of Array Scanning for Image Transmission Systems is two-fold:

- (1) Array scanning has the potential for increasing the throughput rate for image transmission systems that are not channel bandwidth limited.



<u>METHOD OF SPATIAL SAMPLING</u>	<u>METHOD OF SCANNING</u>	<u>ENCODED SIGNAL DESCRIPTION</u>	<u>INHERENT PROCESSING CAPABILITY</u>
a. 		POINT SCANNED	ONE-DIMENSIONAL PROCESSING
b. 		LOCAL NEIGHBORHOOD VECTOR SCANNED	SEPARABLE TWO-DIMENSIONAL LOCAL NEIGHBORHOOD IMAGE PROCESSING
c. 	 OR STEP AND REPEAT	LOCAL NEIGHBORHOOD MATRIX SCANNED	GENERAL NONSEPARABLE TWO-DIMENSIONAL LOCAL NEIGHBORHOOD IMAGE PROCESSING
d. 	NONE	FULL IMAGE SCANNED	GENERAL TWO-DIMENSIONAL IMAGE PROCESSING

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Figure 1.1-1  
An Hierarchy of Image Spatial Sampling Techniques

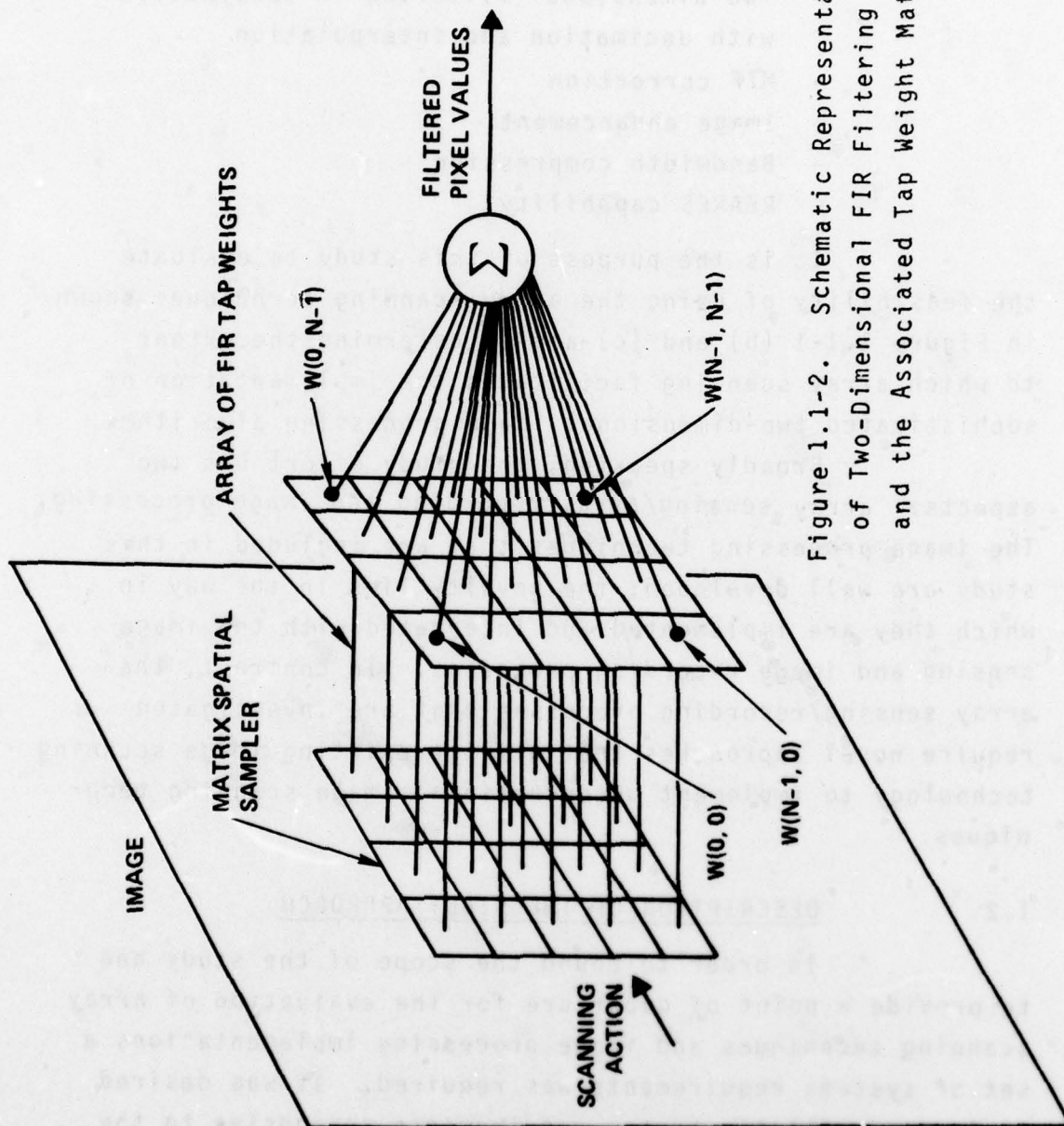


Figure 1.1-2. Schematic Representation of Two-Dimensional FIR Filtering and the Associated Tap Weight Matrix

(2) Array scanning greatly facilitates implementation of certain two-dimensional image processing tasks, including:

- Two-dimensional filtering in conjunction with decimation and interpolation
- MTF correction
- Image enhancement
- Bandwidth compression
- REARCS capability

It is the purpose of this study to evaluate the feasibility of using the array scanning techniques shown in Figure 1.1-1 (b) and (c) and to determine the extent to which array scanning facilitates the implementation of sophisticated two-dimensional image processing algorithms.

Broadly speaking, the study effort has two aspects: array sensing/array recording and image processing. The image processing techniques that are included in this study are well developed; the novelty lies in the way in which they are implemented and integrated with the image sensing and image recording processes. In contrast, the array sensing/recording processes that are investigated require novel approaches that utilize existing image scanning technology to implement unconventional image scanning techniques.

## 1.2 DESCRIPTION OF THE STUDY APPROACH

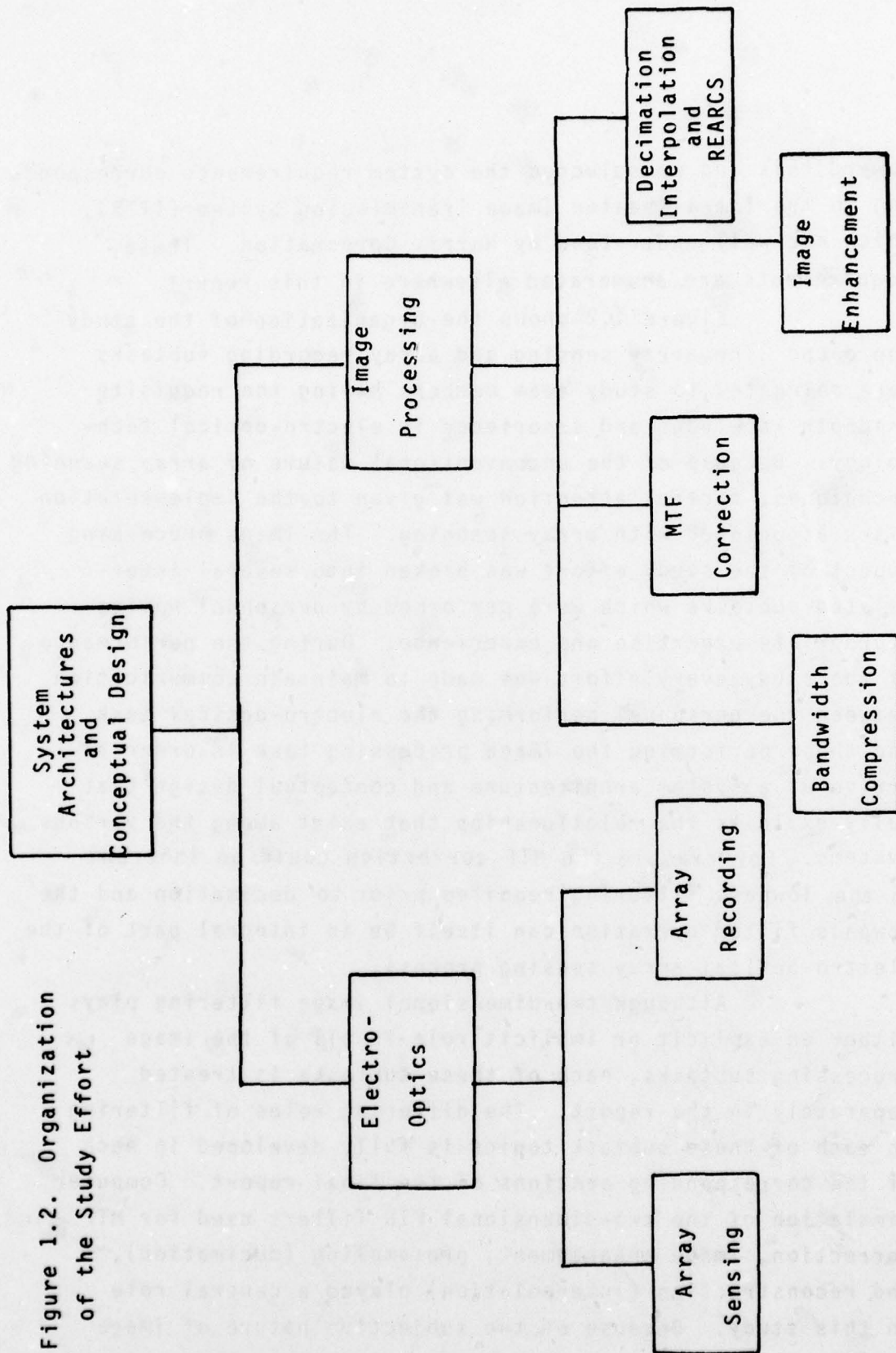
In order to bound the scope of the study and to provide a point of departure for the evaluation of array scanning techniques and image processing implementations a set of systems requirements was required. It was desired to impose realistic system requirements responsive to the current needs of the military regarding image dissemination;



toward this end we selected the system requirements corresponding to the Intra-Theater Image Transmission System (IITS), which are well-understood by Harris Corporation. These requirements are enumerated elsewhere in this report.

Figure 1.2 shows the organization of the study approach. The array sensing and array recording subtasks were delegated to study team members having the requisite in-depth knowledge and experience in electro-optical technology. Because of the unconventional nature of array scanning techniques, careful attention was given to the implementation risks associated with array scanning. The image processing aspect of the study effort was broken into several inter-related subtasks which were performed by personnel having appropriate expertise and experience. During the performance of the study every effort was made to maintain communication between the personnel performing the electro-optical task and those performing the image processing task in order to arrive at a system architecture and conceptual design that fully exploits the relationships that exist among the various systems. For example the MTF correction could be inherent in the lowpass filtering required prior to decimation and the lowpass filter operation can itself be an integral part of the electro-optical array sensing process.

Although two-dimensional image filtering plays either an explicit or implicit role in all of the image processing subtasks, each of these subtasks is treated separately in the report. The differing roles of filtering in each of these subtask topics is fully developed in each of the corresponding sections of the final report. Computer simulation of the two-dimensional FIR filters used for MTF correction, image enhancement, presampling (decimation), and reconstruction (interpolation) played a central role in this study. Because of the subjective nature of image quality evaluation, the computer simulation of the various



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image processing algorithms was of paramount importance in performing the subsystem tradeoffs. The study philosophy regarding the image processing tasks has been to first thoroughly develop the analytical or theoretical aspects of the issues relating to aliasing, decimation, interpolation, MTF correction, and image enhancement, and then use computer simulation to provide the data required to perform subjective evaluation of the system or subsystem tradeoffs. We feel that this approach is appropriate and that it has served the purposes of this study effort well. Wherever appropriate in this report we have included half-tone reproductions of computer processed imagery developed on 3M 7771 heat-processed dry silver photographic paper. The source image used in all the computer simulation experiments was a composite image that we constructed using an aerial photograph containing both natural and man-made features, a quarter of a frequency wedge, and selected typewritten text. This composite image has proved to be quite useful for our experiments--especially the frequency wedge portion since this allowed us to accurately compare both aliasing content (evidenced as moire patterns) and in-band frequency content of the images after various filtering operations. It is worthy of comment that the source image was scanned at 125 lines-per-inch (lpi) whereas the system requirements we assumed for this study mandated 384 lpi as the maximum possible resolution. We chose to use 125 lpi resolution for our experiments because this allowed us to properly perform the subjective evaluation of the various filtering operations while maintaining a manageable computer run time requirement. In spite of this, the run times of some of the experiments were very long and were accomplished on an overnight or over-weekend-run basis. For example, a single 31 x 31 FIR filtering operation without decimation or interpolation required 11 hours and 43 minutes for the 350 x 350 pixel source image we used; the same



source image scanned at 384 lpi would have required 110.5 hours (4.6 days), which is clearly prohibitive in view of the large number of filtering simulations that we felt were necessary.

The decimation/interpolation experiments constituted the bulk of the simulation effort. The decimation/interpolation filters that were simulated included 31 x 31, 15 x 15, and 9 x 9 FIR quarter- and half-bandwidth filters, both separable and nonseparable, with Kaiser window parameter  $\beta = 3, 1, \text{ and } 0$ . This permitted a thorough evaluation of the tradeoffs among stopband attenuation, transition bandwidth, passband ripple, aliasing, inband image detail, and implementation complexity. The MTF correction, image enhancement, and bandwidth compression simulations received less attention because of the limited number of parameter tradeoffs involved. The theoretical and practical limitations of array sensing and array recording techniques were thoroughly examined. The final phase of the study effort draws upon the results and tradeoffs of all the subtasks in order to synthesize several candidate ASITS architectures, whose advantages and disadvantages are discussed and traded off. The result is a final conceptual design whose proposed architecture is evaluated.

### 1.3 ORGANIZATION OF THE STUDY FINAL REPORT

Following the introductory paragraph 1.0, the final report is organized as follows.

Paragraph 2.0 documents the results of the array sensing subtask in which both theoretical and practical limitations on array sensing implementation and performance were explored. The tradeoffs among the various parameters of the array sensing technique are displayed graphically.

Paragraph 3.0 describes the tradeoffs and reports the achievable limits regarding array recording implementation and performance. Again, the tradeoffs are presented graphically in order to facilitate an intuitive understanding of the complex relationships among array recording parameters. Although some of the technologies required for array recording are the same as those required for array sensing, the applications of the technologies are sufficiently different that array sensing and array recording warranted separate paragraphs in the final report.

Paragraph 4.0, Data Collection and Processing, contains all of the theoretical and computer simulation results regarding two-dimensional decimation, interpolation, lowpass filtering, MTF correction, and image enhancement. In addition, the theoretically derived performance limitations resulting from using nonideal or defective CCD's for FIR implementation are reported in this section in the form of computer-generated graphical plots.

Paragraph 5.0 presents a tutorial overview of the bandwidth compression techniques of interest in the ASITS application. The implementation tradeoffs involved for bandwidth compression algorithms using array scanning techniques are reserved for the following paragraph, 6.0.

Paragraph 6.0, Candidate Systems, treats the complexity versus performance tradeoffs for array scanning systems of various complexity. In this paragraph we show top-level implementation diagrams for array scanning systems having complexity ranging from that of minimal capability systems to those having full IITS-type capability.

Paragraph 7.0 details the conceptual design and expected performance for the candidate system of paragraph 6.0 deemed most cost-effective for the ASITS application.

The summary, conclusion, and recommendations of the study effort are contained in paragraph 8.0.

Appendices have been included as deemed appropriate to the presentation of the study results.

#### 1.4

#### CONCLUSIONS AND RECOMMENDATIONS

The results of this study indicate that an array scanning image transmission system is both feasible and advantageous when compared to a conventional single point scanning image transmission system. The study effort focussed upon array scanning implementations and tradeoffs and the implementation of sophisticated processing algorithms using the CCD technology that is so well suited to array scanning techniques. It was found that the state-of-the-art of electro-optics technology supports one-dimensional sensing and recording arrays and that two-dimensional sensing arrays can be easily synthesized from one-dimensional sensing arrays. After evaluating many candidate systems, we present a conceptual design for a sophisticated ASITS whose implementation is built entirely around two commercially available CCD's. Based upon the results of this study, we recommend a multiphase follow-on study effort. Section 8.0 contains a more detailed description of the study conclusions and recommendations.



## 2.0 ARRAY SENSING

The function of a scanning system is to extract data from a source image and prepare it for transmission. This usually implies that a minimum configuration will consist of a scanning method, an illumination source, and a detection system, with appropriate electronic control functions synchronizing their operation. Requiring the system to be capable of array scanning places an additional demand on the optical system: either the illumination source or the optical receiver (or both) must be capable of distinguishing a number of different positions for each scan location. In addition, the requirement to resolve simultaneously a number of pixels can place restrictions on the nature of the scanning technique, because of the larger apertures required and the size of the array devices themselves.

There are a number of techniques that can be used to provide multichannel or multidimensional array scanning capabilities. These include acousto-optic devices, fiber-optic arrays, photodiode detector arrays, and LED arrays. Each of these approaches can be combined with conventional scanning techniques to provide an array scanning/image dissection system. Among conventional scanning techniques, the most successful and generally useful for tactical imagery applications are galvanometer scanners and rotating drum scanners. Galvo scanners, which move only the optical beam in relation to the stationary document, can provide the significantly greater speeds required in some applications. Drum scanners, however, are particularly relevant to array scanning applications, since their use of stationary sources and detectors can support more complex optical configurations as

required in many array techniques.

In this section we will provide a discussion of the tradeoffs and implementation considerations associated with currently accessible array technologies, including both basic introductions to the capabilities and limitations of specific components, and conceptual system descriptions of some viable array scanning systems. After several systems have been described briefly, some more specific goals (as to formats, data rates, etc.) will be assumed and used as the basis for some detailed tradeoff considerations of a specific array scanning system design. The section will conclude with some discussion of alternative approaches to meeting the specific goals of the main scanner system.

Emphasis in this section will be placed on the scanning of one-dimensional linear arrays, in preference to the block matrix approach. There are two reasons for this emphasis: first, the linear devices currently have significant cost and implementation advantages over matrix-oriented systems, often without sacrificing much in the way of performance; second, the development of two-dimensional input/output devices is not yet adequate to support the multiple, field-operational units which would be required for a useful imagery transmission network.

Finally, it should be noted that many of the devices and techniques discussed in this section are also relevant to array recording; thus additional data and tradeoff discussions on these components may be found in Section 3.0.

## 2.1 COMPONENTS FOR ARRAY SCANNING

A necessary precondition to an understanding of the array scanning system approaches to be discussed in this section (and the reporting techniques of the following section) is a familiarity with the electro-optical components which compose these systems. In this subsection we will present brief introductions to each key concept and technique to be used in the subsequent system descriptions.

### 2.1.1 Scanning Techniques

Since very large format area sensing devices are not currently available (or on the horizon), a means is required to scan the operational area of a smaller sensor across the width of the image to be transmitted. For most image transmission systems, drum and galvanometer based scanners have proven to be the most successful; also, in small-format, high-speed applications, acousto-optic devices have many attractive features, especially their speed capabilities and the absence of moving parts. These three types of scanning devices will be used extensively in the system descriptions to follow; accordingly a brief summary of their nature and capabilities is appropriate here.

#### 2.1.1.1 Galvanometer Scanners

A galvo scanner is a moderate-speed, random access device based on the motion of a coil within a changing magnetic field. One end of the coil armature shaft supports a thin mirror which moves with the shaft. Thus, electrical inputs to the driving field can control mirror position and hence beam angle. A simple illustration of this situation is shown in Figure 2.1.1.1.



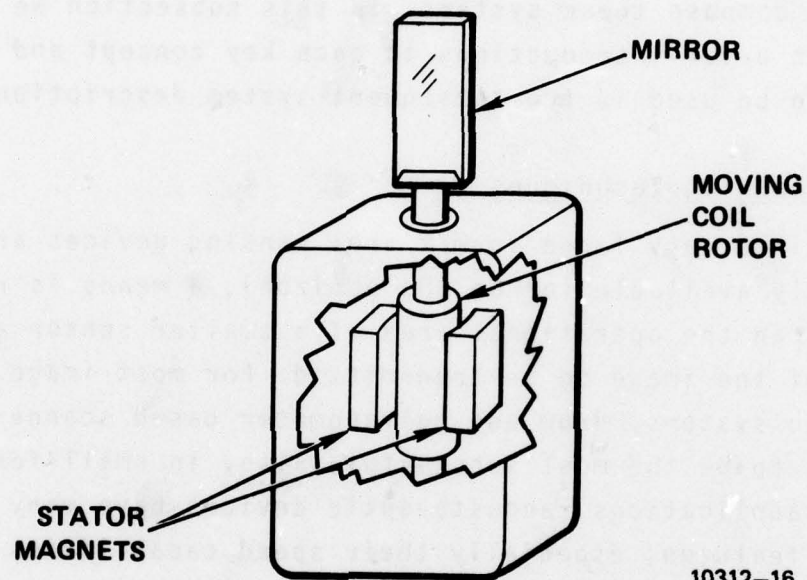


Figure 2.1.1.1  
Basic Moving - Coil Galvanometer Structure

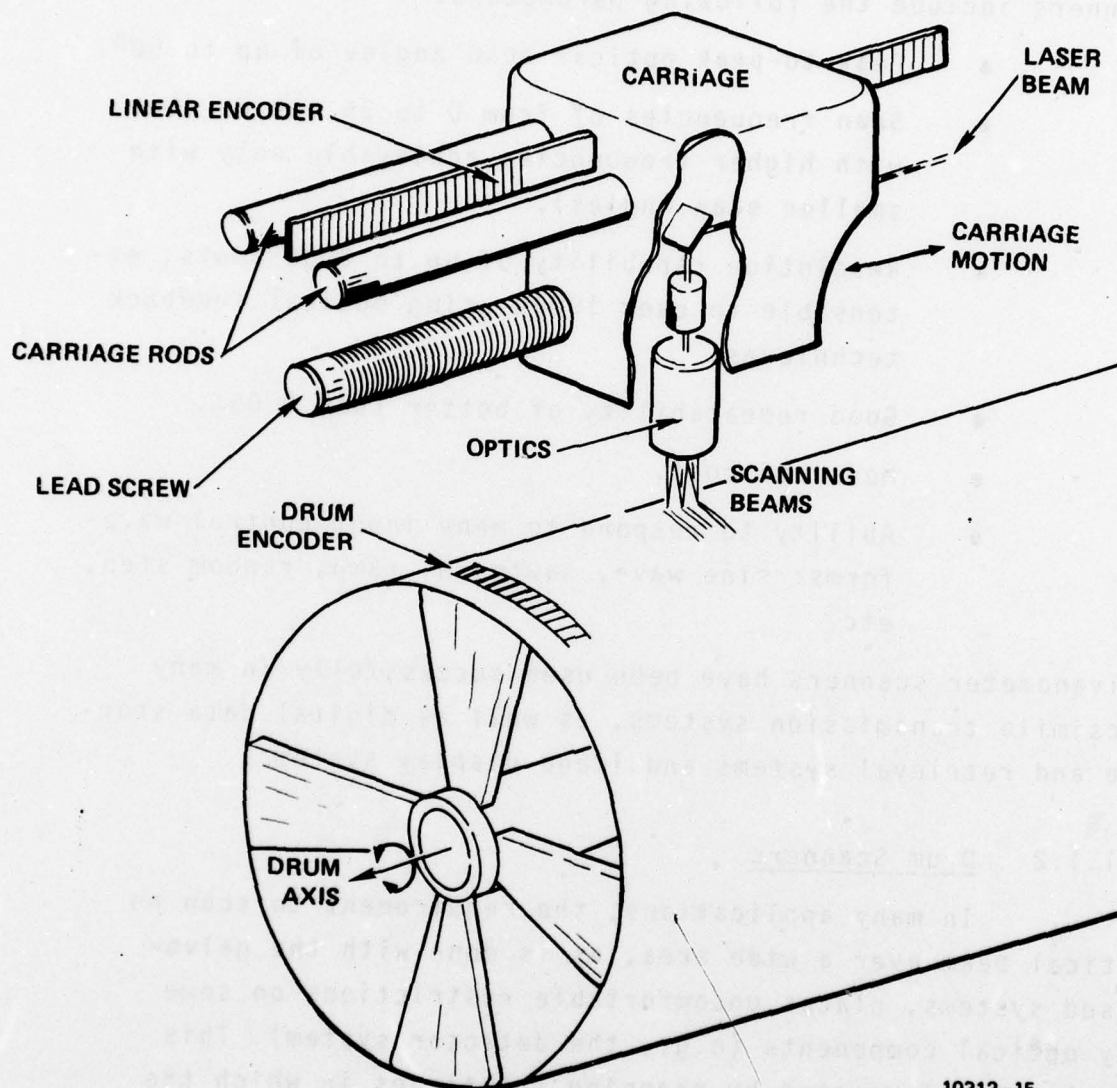
Some of the capabilities and limitations of galvo scanners include the following parameters.

- Peak-to-peak optical scan angles of up to 60°.
- Scan frequencies of from 0 to 25,000 Hz (but with higher frequencies achievable only with smaller scan angles).
- Resolution capability of up to 2000 spots, extensible to over 10000 using optical feedback techniques.
- Good repeatability of better than 0.05%.
- Moderate cost.
- Ability to respond to many input control waveforms: sine wave, sawtooth, ramp, random step, etc.

Galvanometer scanners have been used successfully in many facsimile transmission systems, as well as digital data storage and retrieval systems and image display systems.

#### 2.1.1.2 Drum Scanners

In many applications, the requirement to scan an optical beam over a wide area, as is done with the galvo-based systems, places uncomfortable restrictions on some key optical components (e.g., the detector system). This problem is eliminated by scanning techniques in which the scanned medium is moved past a stationary optical beam in one of the two dimensions; the major technique in this category is the drum scanner. Such a system is illustrated in simplified form in Figure 2.1.1.2. The optical beam



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Figure 2.1.1.2  
Multi-Beam Drum Scanning Configuration



(which for an array scanning may be a multiple beam) is translated parallel to the drum axis by a moving optical head, while the drum rotates the scanned document perpendicular to the beam's motion, creating an X-Y raster scan of the drum surface. The advantage of such a system is that, since the beam-forming optics do not have to participate in satisfying a wide-area scanning constraint, they are better able to produce a complex beam structure, as would be required in an array scanning application. Also, since the beam/surface interaction area is small, a smaller detector can be used, permitting higher speed devices and/or higher sensitivities and consequently reduced input light level requirements.

The capabilities of drum scanners currently available include the following:

- Capacity for fairly large documents, up to ~14 inch length.
- Speeds up to about 1800 RPM.
- Resolution of up to 500-700 lines per inch.
- Good reliability compared to more complex scanning mechanisms.

Applications of drum scanners have included primarily facsimile image transmission systems; a few data storage and retrieval systems have also been drum-based.

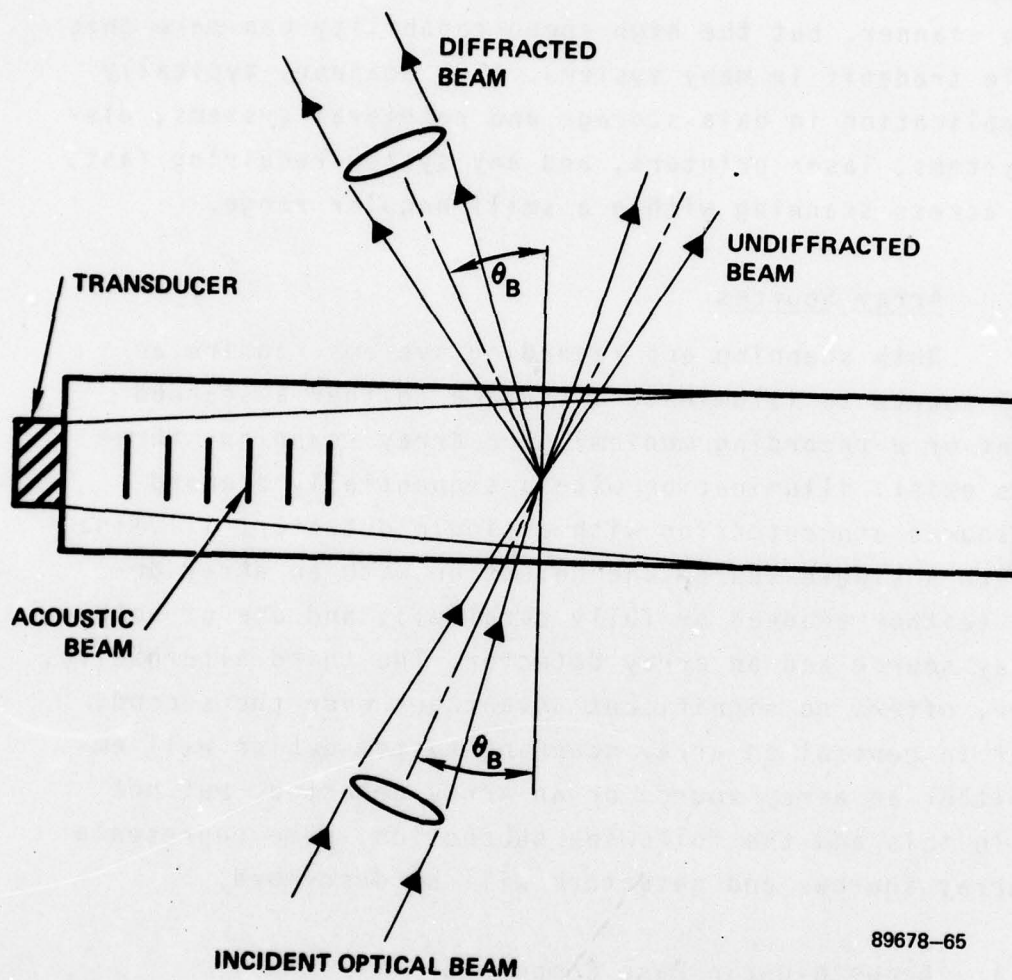
#### 2.1.1.3 Acousto-Optic Devices

In many applications requiring fast, small field scanning, acousto-optic devices are useful. An acousto-optic device is a block of transparent material (various types of

glass, for example) through which we pass the laser beam that we want to modulate or deflect. In the crystal, the light beam encounters the acoustic wave. These devices typically operate in the "Bragg" diffraction mode, which implies that they behave like thick gratings rather than thin diffraction gratings. This mode produces two main output beams as shown in Figure 2.1.1.3. One beam, the undiffracted component, has not interacted with the acoustic wave and is usually discarded. The other beam, the diffracted component, has interacted with the acoustic wave. Temporal modulation of the acoustic wave amplitude can be used to temporally modulate the amplitude of the diffracted light wave; more importantly for scanning applications, frequency modulation of the acoustic wave will produce angular variation of the light wave, since the diffraction angle is proportional to the spatial frequency of the acoustic wave. The acoustic wave is introduced into the crystal by means of an electromechanical transducer which is bonded to one face of the crystal, and which in turn is connected to an appropriate source of RF electrical energy and associated control circuits for amplitude or frequency modulation.

When used to produce angular scanning of a beam, an acousto-optic device is known as an acousto-optic beam deflector (AOBD). Some of the pertinent capabilities of AOBD scanners include the following:

- Peak-to-peak optical scan angles of up to  $2^{\circ}$ .
- Scan frequencies greater than 250 kHz.
- Up to 2000 resolvable spots.
- Capability to respond to a variety of waveforms: sine wave, sawtooth, ramp, random step, etc.



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Figure 2.1.1.3  
Details of Acousto-Optic Interaction



This type of scanner is usually somewhat more costly than a galvo scanner, but the high speed capability can make this a viable tradeoff in many systems. AOBBD scanners typically find application in data storage and retrieval systems, display systems, laser printers, and any system requiring fast, random access scanning within a small angular range.

#### 2.1.2 Array Sources

Both scanning and recording systems require an optical source to illuminate a surface (either a scanned document or a recording medium). For array scanning, three options exist: illumination with a sequentially scanned array source and detection with a single detector; illumination with a single source and detection with an array detector (either scanned or fully parallel); and use of both an array source and an array detector. The third alternative, however, offers no significant advantages over the second, so that in general an array scanning system design will employ either an array source or an array detector, but not both. In this and the following subsection, some representative array sources and detectors will be described.

##### 2.1.2.1 Acousto-Optic Page Composers

In the previous subsection, a description of the functioning of acousto-optic devices was given, and it was noted that, if the drive frequency is held constant, such a device can be used as a modulator simply by varying the amplitude of the input signal. When an array of such modulating transducers is placed on a single crystal to form a modulating array, the resultant device is known as an acousto-

optic page composer (AOPC). A schematic drawing of an AOPC is shown in Figure 2.1.2.1. When illuminated by a line of laser light, the diffracted beams from the AOPC's transducers form a fully parallel, high performance linear array source. AOPC's with up to 128 channels have been fabricated, providing total data throughput rates of up to 780 Mbits/s. The specifications of a 32-channel device are given in Table 2.1.2.1, and provide a representative example of the capabilities of this type of array source.

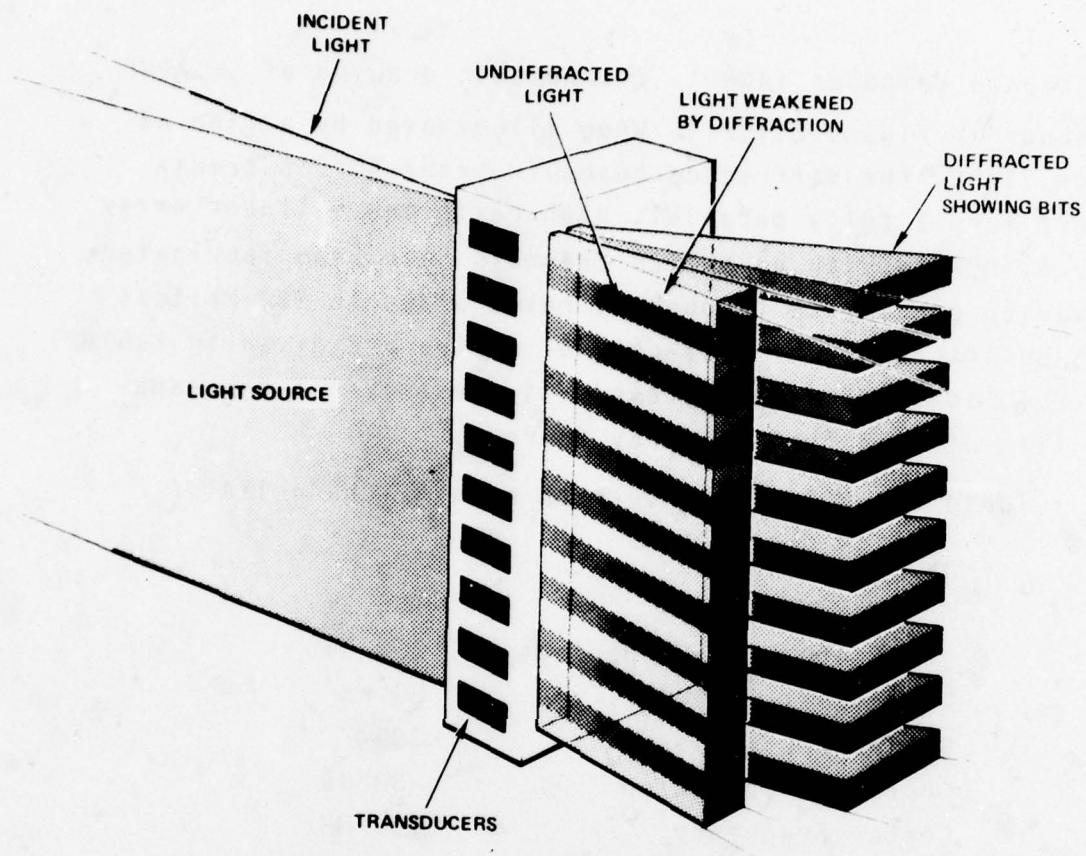
Table 2.1.2.1. Specification of a 32-Channel AOPC

No. of Channels	32
Transducer Size	100 $\mu\text{m}$
Center-to-Center Spacings	250 $\mu\text{m}$
Rise Time (10-90%)	20 ns
Beam Use Efficiency	30%
Crosstalk Isolation	-33 dB
Center Frequency	150 MHz
Material	TeO <sub>2</sub>

Because of their high speed capability and potential for operation with high power lasers, AOPC's are used in a wide variety of optical storage and processing applications, including holographic recorders, direct spot recorders, data processors, display systems, etc.

#### 2.1.2.2 Light-Emitting Diode Arrays

A semiconductor diode in which a proper energy gap exists between the valence band and the conduction band in the P-type material can be made to emit light when biased in the forward direction. The wavelength emitted depends



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Figure 2.1.2.1  
Acousto-Optic Page Composer



on the energy gap width, and the efficiency of the device is controlled by the tendency of the electrons in the material to emit photons rather than phonons (vibrations) as they fall from the excited state. Light emitting diodes (LED's) are available in most areas of the visible spectrum, as well as the near infrared. Both single element and array sources have been constructed and delivered, with good reliability and flexibility of configuration.

LED's can be used in any optical system in which a relatively low-power, incoherent source is acceptable. Linear array devices have been constructed with some or all of the following characteristics:

- Output wavelength: 660 nm
- Output power: 1  $\mu$ W/mA drive
- Maximum output: 400  $\mu$ W pulsed, 40  $\mu$ W continuous
- Number of elements: hundreds
- Array output uniformity:  $\pm 30\%$
- Rise time: 50 ns
- Commutation rate: 200 kHz

Additionally, progress in element spacings is continuing, with 7.5 mils achieved and 2 mils projected for the near future. This combination of capabilities make LED arrays potentially attractive as a source for the array scanning application.

LED arrays have chiefly found use in document scanners, film annotation, alpha-numeric readout, pattern recognition and display systems.

#### 2.1.2.3 AOBD Techniques

The acousto-optic beam deflector, or AOBD, as described above in Paragraph 2.1.1.3, is a convenient means of achieving rapid, narrow-angle scanning. Another potential use of such a device is in the production of multiple beams from a single beam. By inputting simultaneously several drive frequencies, a number of acoustic frequencies can be produced within the crystal, each frequency capable of deflecting some light in a different direction. In this case, the frequencies are held constant (so that no angular motion of the beams takes place), and the beams can be individually intensity modulated by controlling the strength of each input frequency. This technique will be described in more detail in Paragraph 3.0, where it will be used as the basis of an array recording system.

An additional AOBD technique, in which the various beams are not all present simultaneously, is not a true array technique, but will be of importance for the system design to be presented in a later section of this paragraph; this technique will be described in sections 2.1.4 and 2.3.3.2.

#### 2.1.3 ARRAY DETECTORS

As previously described, one viable option for the production of array information involves the use of a scanned array source and a single detector; in that case, an array detector is not used. In the other possible systems, however, scanned or parallel array detectors are necessary. The two most useful of the current devices in this category are fiber-optic/discrete detector arrays, and scanned integrated detector arrays.

#### 2.1.3.1 Fiber-Optic/Discrete Detector Arrays

One effective way to acquire array information at high speeds is through the use of an array of optical fibers and discrete detectors, as shown in Figure 2.1.3.1. By imaging the linear area to be scanned onto the input surface of the fiber-optic linear bundle, information from each pixel location can be conducted to a discrete detector for amplification and processing. The use of discrete detectors permits advanced integral preamplifier designs to be used and thus increases the speed and SNR achievable by the system. Fiber arrays with several hundred elements can be constructed; some other readily achievable array parameters are as follows:

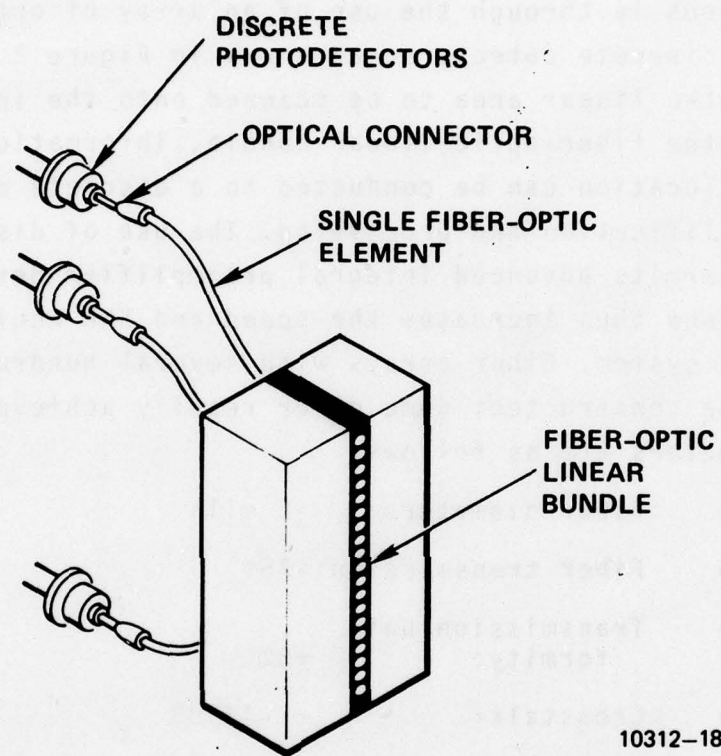
- Fiber diameters:  $\geq 5$  mils
- Fiber transmission:  $> 75\%$
- Transmission uniformity:  $\pm 5\%$
- Crosstalk:  $< -45$  dB
- Per-channel data rate: 10 Mb/s

Typical applications of fiber-optic/discrete detector arrays include high-speed optical data storage and retrieval systems and document scanners.

#### 2.1.3.2 Integrated Detector Arrays

Although not as fast or as adaptable to parallel output systems as the discrete approaches, integrated detector arrays are another important option for array detection. These devices include charge-coupled devices (CCD's),





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Figure 2.1.3.1  
A Linear Fiber-Optic Array

charge-injected devices (CID's), phototransistor arrays (PTA's) and others. The primary advantage of such devices is that they can accommodate both a number of photosensitive areas and a preamplification stage on the same substrate; this makes possible low-noise amplification of the received signals and thus enhances the SNR and speed capabilities of the units. The space problems associated with placing multiple amplifiers on the substrate with the photosensors make the scanned approach much more common. In this approach, signals from the photosites are serially gated through a common amplifier prior to leaving the chip. Parallel-output integrated photosensitive arrays are less common, and generally provide poorer SNR performance because of the off-chip signal amplification; however, such devices do exist, and are capable of much greater information throughput rates than are the serially scanned arrays.

The following table illustrates the capabilities of two typical, serially scanned, wide field integrated photo sensors.

Table 2.1.3.2. Integrated Photo Sensor Characteristics

	Device 1 (Fairchild)	Device 2 (Reticon)
Array Size	1728X1	1024X1
Cell Size	0.51X0.67 mils	1 X 1 mil
Cell Center Spacings	0.51 mils	1 mil
Operational Method	CCD buried channel	Silicon gate MOS
Dynamic Range	500:1 @ 1 MHz	1000:1
Max. Clock Rate	1 MHz	10 MHz
Saturation Exposure	0.5 $\mu\text{J}/\text{cm}^2$	0.11 $\mu\text{J}/\text{cm}^2$

Applications of these devices include page scanning, facsimile readers, optical character recognition, process control, film reading, and spectroscopy. Self-scanned integrated photosensors are also available as matrix arrays in sizes up to about 400 x 400.

#### 2.1.4 PSEUDO-ARRAY GENERATION

To achieve the full speed capabilities of array scanning, parallel generation of the picture elements is required. However, in many cases, such as when system bandwidth is limited by a data transmission link, speed is not the prime concern, and other methods of obtaining information for array processing may be considered. In particular, if the data associated with an array of picture elements can be generated serially and stored in analog form, parallel processing can then be done as if they were generated simultaneously.

One approach to achieving the required array information is to buffer entire scan lines until enough lines are stored to cover the matrix of interest. This approach is straightforward, but requires (for all but the smallest scan formats) an uncomfortably large buffer, e.g.,  $N$  lines times several thousand pixels for arrays of size  $N$  by  $N$ . Another approach, which will be discussed in detail in Section 2.3.3 below, involves vertical dithering of a horizontal raster, so that the  $N^2$  pixels of a matrix array are scanned out contiguously, rather than  $N$  in each full line.

In both approaches, a method of storage of information is required, to allow serial scanning and parallel processing both to be accomplished. When full lines of data



must be buffered, digital storage is currently the only practical approach. For the smaller, dithered raster scan technique, however, analog storage is feasible. Two techniques which may be used to accomplish such analog storage are an electrical tapped analog delay line (TAD) and an acousto-optic delay line (AODL).

#### 2.1.4.1 Electrical Delay Line

Storage and shifting of analog values has become economically viable with the recent introduction (primarily by Reticon, Inc.) of TAD devices in integrated form. In these devices, analog values are represented by quantities of charge stored on capacitive nodes in a transistor gate network. By adjusting the potentials in the semiconductor material properly the charge can be shifted from node to node at each clock pulse. Also, taps at each node make the values available in parallel to the outside world for simultaneous processing. These TAD devices can achieve clock rates of 5 MHz and can have up to 32 stages (much longer devices have been built, but without the feature of parallel output taps). Additional discussion of TAD devices and test results of a particular unit are provided in section 2.3.3.3.3.

#### 2.1.4.2 Acousto-Optic Delay Line

Electrical TAD devices provide a flexible, integrated package, but have limitations in the areas of pulse fidelity and number of output taps. By using an acousto-optic device, some of these problems can be overcome. The technique is a standard one in acousto-optics: modulating

the analog samples onto an RF carrier and inputting them to an acousto-optic crystal via a piezo-electric transducer. As the pressure waves, whose intensities represent the analog values, propagate down the crystal, they are available for simultaneous illumination by a laser source, and hence parallel processing. The primary limitations of such a system are the available crystal sizes and the F/numbers in the optical system itself. Additional details on acousto-optic devices are provided in sections 2.1.1.3 and 2.3.3.3.1, while the use of the AODL in serial-to-parallel conversion is central to the system described in section 2.4.2.

#### 2.1.5 Direct Electronic Fourier Transformation

Another type of component which can be of significant use in array scanning systems is the direct electronic Fourier transform (DEFT) device. This is a device which can produce as an output the Fourier transform of an image projected on its surface; thus it provides the capability for a variety of two-dimensional filtering and bandwidth compression operations with minimal hardware complexity. Because of these filtering applications, the DEFT device could validly be described in the section of this report dealing with filtering and bandwidth compression; by placing it here, however, we can more easily see its relationship to some other array sensing techniques, as well as its usefulness for array scanning.

There has been considerable effort expended since 1972 toward the development of a "Fourier Transform Camera", which would generate a set of time-sequential electrical signals corresponding to the Fourier Transform of an image projected upon it. This effort has been supported at

Syracuse University by RADC under Contract numbers F30602-72-C-0403 and F30602-73-C-0277. The results of this research are contained in two final reports:

Direct Electronic Fourier Transforms of Optical Images, RADC-TR-73-254, August 1973<sup>[2]</sup> and

DEFT: Progress on Direct Electronic Fourier Transforms of Two-Dimensional Images, RADC-TR-74-244, October 1974<sup>[3]</sup>.

Additional research on this topic has been funded by the U.S. Army Night Vision Laboratory, Ft. Belvoir, Va., again at Syracuse University, under Contract No. DAAG53-76-C-0162. An interim report for the period 10 June 1976, through 9 June 1977 is entitled Two-Dimensional Direct Electronic Fourier Transform (DEFT) Devices: Analysis, Fabrication, and Evaluation<sup>[4]</sup>.

In this section we trace the progress of this research as reported in the above reports and evaluate the applicability of present state-of-the-art DEFT devices to an image compression and transmission system.

The research reported in [2] deals with theoretical aspects of two-dimensional surface acoustics, image reconstruction, and motion detection, and describes the experimental aspects of the design and construction of a one-dimensional CdS surface acoustic wave (SAW) "camera".

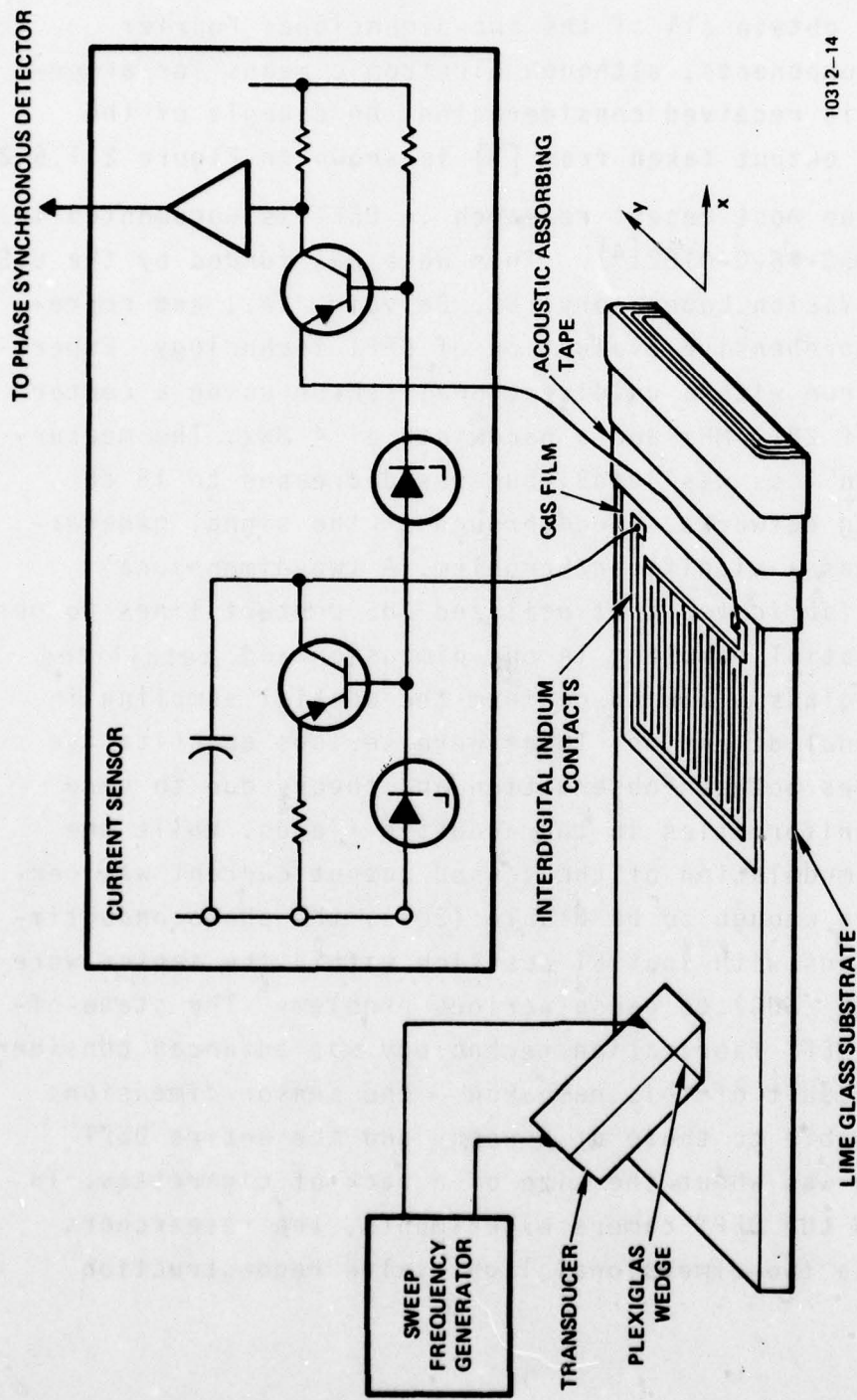
The SAW camera utilizes the elastophotoconductivity of the semi-conducting CdS film deposited onto a glass substrate to correlate the impinging light pattern with the SAW launched into the glass substrate by the transducer.

An electrical current proportional to the photoconductivity of the CdS film is sensed phase-synchronously



to provide a voltage output corresponding to the particular Fourier spatial frequency component. The spatial frequency of this Fourier Transform component is related to the sinusoidal frequency driving the electrical transducer through the velocity of propagation of the surface strain wave in the substrate. Different Fourier spatial frequency components can be obtained by varying the wavelength of this surface strain wave. Sweeping the frequency of the generator driving the electrical transducer provides a time-serial output sequence of the various Fourier transform spatial frequency components of the incident light pattern. A one-dimensional DEFT camera was fabricated and tested for several simple test illumination patterns consisting of alternating strips of light and dark. The experimental results demonstrated the validity of the approach for the one-dimensional SAW camera. The remainder of the report dealt with various theoretical aspects concerning the two-dimensional DEFT camera.

The research for the following year was reported in RADC-TR-74-244<sup>[3]</sup>, and described the experimental results obtained using a two-dimensional DEFT camera. The experimental setup that was used is shown in Figure 2.1.5-1 taken from [3]. The sensors used exhibited improved sensitivity and resolution compared to previous devices. The two-dimensional DEFT camera utilized 100 sensors in parallel which were fabricated by deposition of interdigital contacts on a square CdS film. This DEFT camera was used in extensive focus detection and motion detection experiments with encouraging results. Also reported was the building and testing of an elastobirefringent light valve for one-dimensional



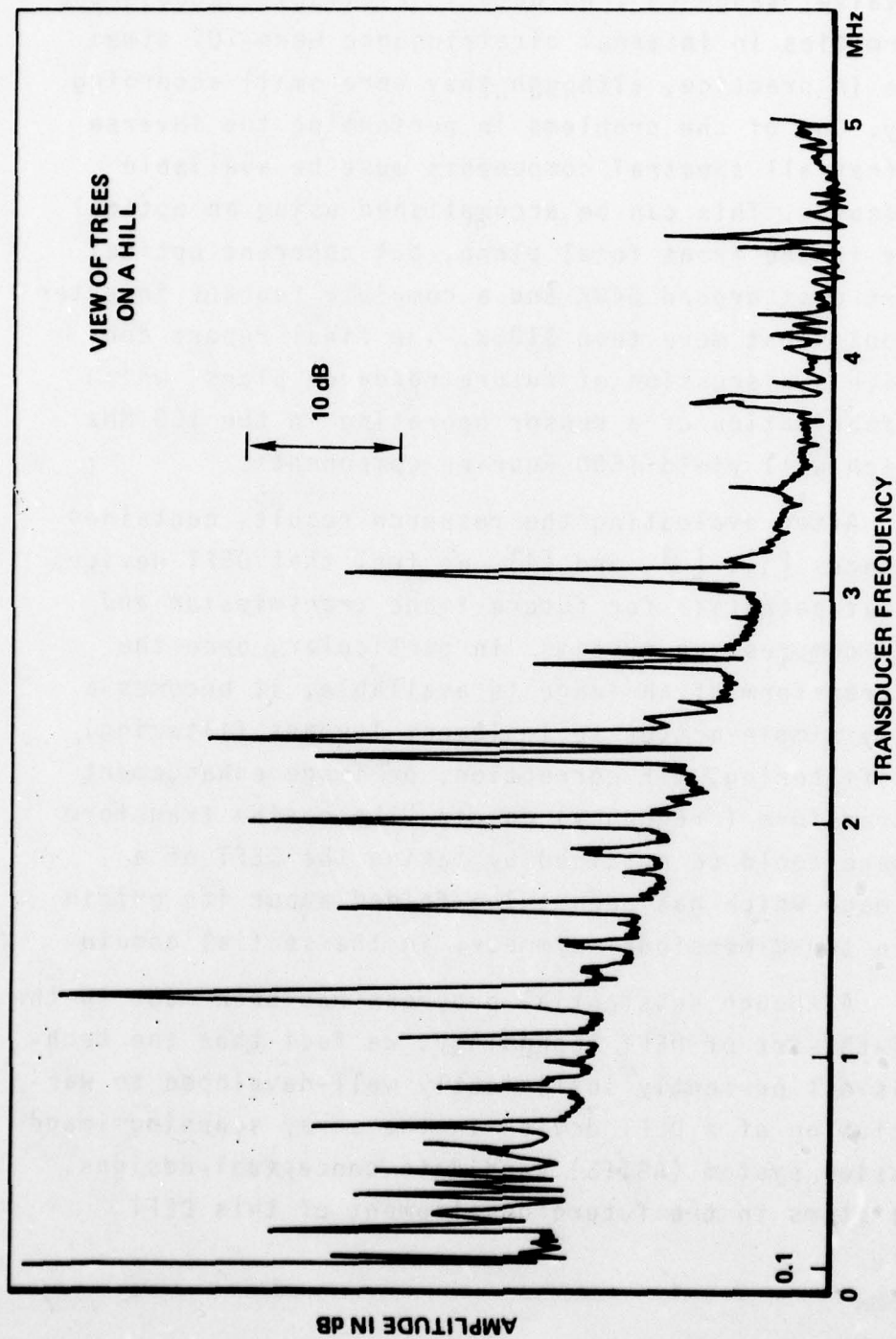
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Figure 2.1.5-1  
Two-Dimensional Interdigitated Elastophotovoltaic Sensor

reconstruction. In the work reported for this phase of the research, mechanical rotation of the image was required in order to obtain all of the two-dimensional Fourier transform components, although electronic means for accomplishing this received consideration. An example of the DEFT camera output taken from [3] is shown in Figure 2.1.5-2.

The most recent research in DEFT is documented in Report DAAG53-76-C-0162[4]. This work was funded by the U.S. Army Night Vision Laboratory, Ft. Belvoir, Va., and represents a comprehensive evaluation of DEFT technology. Experiments were run with a unidirectional sensor using a center frequency of 28.2 MHz and a bandwidth of 4 MHz. The measured insertion loss was 40 dB, but was decreased to 18 dB via matching networks. Feedthrough of the signal generator source was a significant problem. A two-dimensional sensor was fabricated that utilized CdS contact lines to perform the spatial sampling in one dimension and metallized lines on a glass plate to perform the spatial sampling in the orthogonal dimension. There were serious quantitative discrepancies between observation and theory due to very strong nonuniformities in the acoustic fields. While the AO-induced modulation of the sensed output current was certainly large enough to be usable (20%), the photoconductivity variations with spatial position within the sensor were large enough (50%) to cause serious problems. The state-of-the-art of DEFT fabrication technology was advanced considerably as a result of this research -- the sensor dimensions were comparable to those of a penny and the entire DEFT camera size was about the size of a pack of cigarettes. In addition to the DEFT camera experiments, the researchers fabricated a two-dimensional light valve reconstruction





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Figure 2.1.5-2  
Output Magnitude From a High-Sensitivity DEFT Sensor.  
Overcast Outdoor Scene.

device that relied upon strain-induced birefringence in a light valve "sandwich". No data is available because the nonuniformities in internal birefringence were 100 times too large in practice, although they were small according to theory. One of the problems in performing the inverse DEFT is that all spectral components must be available simultaneously. This can be accomplished using an optical modulator in the front focal plane, but coherent optical modulators cost around \$40K and a complete Fourier inverter system would cost more than \$100K. The final report concludes with a discussion of future research plans, which include fabrication of a sensor operating in the 100 MHz range which will yield 1500 Fourier components.

After evaluating the research results contained in references [2], [3], and [4], we feel that DEFT devices offer great potential for future image transmission and bandwidth compression systems. In particular, once the Fourier transform of an image is available, it becomes a relatively simple matter to implement lowpass filtering, bandpass filtering, MTF correction, or image enhancement in the transform (frequency) domain. The cosine transform of an image could be obtained by taking the DEFT of a source image which has been twice-folded about its origin to obtain two-dimensional symmetry in the spatial domain.

Although substantial progress has been made in the state-of-the-art of DEFT technology, we feel that the technology is not presently sufficiently well-developed to warrant inclusion of a DEFT device in the array scanning image transmission system (ASITS) candidate conceptual designs. Critical items in the future development of this DEFT

technology appear to be increased uniformity of the Cds film spatial responsivity and the fabrication of a usable reconstruction device at a reasonable cost.

## 2.2            ARRAY SCANNING SYSTEM CONCEPTUAL DESIGNS

Having established as a background the nature and capabilities of many of the components which can be used in an array scanning system, we now turn our attention to the tradeoffs that are encountered in designing a system using these components. As a method of covering this subject, we will give brief descriptions of several array scanning systems which can be configured from some of the components described above, describing as we go the advantages and disadvantages of each type of system. In the absence of any specific parametric design goals, only fairly qualitative and conceptual design choices can be made; this will, however, allow adequate presentation of the features of each system, and will form a basis for the following section, where specific design goals will be adopted and more quantitative tradeoff studies described.

### 2.2.1            Drum-Scan/LED System

The details of the optical receiving head of an array scanner based on the combination of Light-Emitting diodes (LED) with drum scanning techniques are given in Figure 2.2.1. The LED array is a linear array of from 5 to 50 emitters aligned transverse to the plane of the figure.



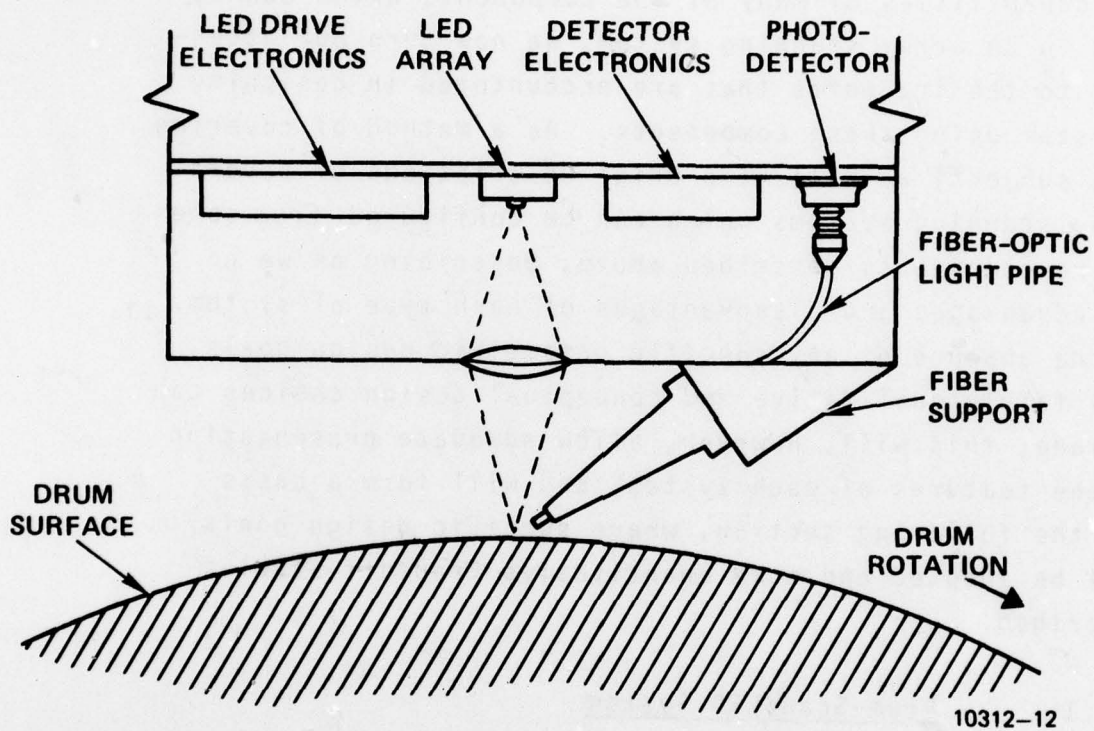


Figure 2.2.1  
Details of Optical Scanner Head for  
Drum-Scan/LED System

The diodes are imaged to the surface of the drum, which supports the document to be scanned. For most systems of interest, the resolution will be high enough and the number of elements small enough that a single light-conducting fiber-optic element can be used to receive the light reflected from the document's surface, independent of which diode is illuminated. This means that a single, fast detector can be used, greatly simplifying the system structure and cost.

To produce information using this system, the LED's are sequentially scanned as the drum rotates. The scan rate of the diodes is sufficiently greater than that of the drum that a vertical line of pixels is scanned out while the drum moves laterally one pixel distance. Thus for an LED array length of  $N$ , each  $N$ -pixel distance traversed by the drum completes a new  $N \times N$  matrix pseudo-array in the manner described in Sections 2.1.4 and 2.3.3.2. Some form of analog stroage for the  $N^2$  pixels must be provided, as shown in Section 2.1.4; however, details of the storage mechanism will not be discussed here, except as needed to assure compatibility with system data rates. The additional components required in the scan head are the commutation electronics for the LED array, and the biasing and post-amplification circuitry associated with the photo-detector.

To determine the performance capabilities of the described system, operating parameters will be chosen which are well within the current state of the various arts involved, thus assuring that the resulting configuration is feasible. The starting point will be the resolution and light budget. For small arrays, LEDs on 2 mil centers are possible, so we will take the required resolution to be

400 lines (pixels)/inch. Thus an array of 1.5 mil square LEDs on 2.5 mil centers would be required for use with a one-to-one imaging system (for maximum light collection efficiency). If we assume an array size of 20 x 20, then the length of the scanned array will be about 50 mils or 1.27 mm. Thus a standard 1.25 mm diameter light pipe can be used with good collection efficiency for the whole array. First, consider the output capabilities of the LEDs: a diode of about 2 mil<sup>2</sup> area can be safely driven at up to 100 mA on a pulsed basis; furthermore, about 1  $\mu$ W of optical output will be generated for each 1 mA of current input; thus about 100  $\mu$ W of signal (at a wavelength of 660 nm) would be produced from a typical LED at the 5% duty cycle envisioned for a 20-element array.

Next, we consider the lens and fiber collection efficiencies. The minimum feasible F/# of the imaging lens is around F/2. This means its efficiency can be as high as about 3%, as given by

$$\eta = \frac{\Omega}{2\pi} = 1 - \frac{2F\#}{\sqrt{1+(2F\#)^2}} = .0298 \quad (2-1)$$

If the collection fiber is located 75 mils from the pixel center, then it will be an F/1.5 system and collect about 5% of the scattered light. Allowing an additional 3 dB of loss for the combined effects of paper reflectivity off-axis collection, and fiber-detector interface loss, we find an expected signal at the detector of approximately 75 nW. For a detector such as the RCA C30914E (an avalanche photo-diode with integral light pipe) this would produce about 10 mV of signal, and the detector NEP of about 0.1 pw  $\sqrt{\text{Hz}}$  would provide adequate SNR up to around 500 KHz,



which is consistent with a system pixel transmission rate of up to 25 KHz. Furthermore, this rate is compatible with the capabilities of the analog storage devices described in Section 2.1.4. Since this analysis is very preliminary, and does not include the important noise considerations involved in the design of the circuitry required to process the detector output, it may be somewhat optimistic; nevertheless, the utility of such a system design for a rather broad class of array scanning applications (perhaps even up to the requirements of the IITS system) has been clearly demonstrated.

A summary of the parameters of the system just described is provided in Table 2.2.1.

Table 2.2.1. Parameters for Drum-Scanned LED Array System

Number of Elements	20
Array Size	20 x 20
LED Size (mils)	1.5
LED Centers (mils)	2.5
LED Commutation Rate (kHz)	500
Resolution (pixels/inch)	400
Pixel Transmission Rate (kHz)	25
Drum Speed (r/min)	180
SNR (Detector Only) (dB)	30

The chief advantage of this system is its compactness, reliability, and the simplicity of its optical design. This is a result of the use of the drum scan technique, which eliminates the wide optical scan/detection require-

ments of other systems, as well as the limitation of system speed, which makes the low power/efficiency of the LED/fiber collector acceptable. For higher speed systems, the parallel data reception scheme of the following subsection may be more appropriate.

#### 2.2.2 Drum-Scan Fiber-Optic Array System

Another drum-based system concept that is appropriate for consideration in medium to high speed array scanning applications is that of Figure 2.2.2. In this system, which superficially resembles that of the previous section, several important changes have been made in pursuit of a specific goal: to shift the burden of pixel separation from the light source to the detector. By passing a laser beam (from a laser mounted external to the scan head) through a series of shaping optics, a line source of light can be produced to illuminate the scanned document at the drum's surface. Then the scattered light is used to image the line onto a linear array of fiber-optic elements, each of which conducts the light from a different pixel to a discrete detector for detection and processing. In the previous system, any non-uniformities caused by the optical configuration could be eliminated by varying the drive to the various LEDs in a compensatory manner; in this system, the analogous procedure (to handle the truncated Gaussian profile of the laser beam, for example) is simply to vary the sensitivity of the individual photodetectors such that a uniform input document results in a uniform output.

To analyze the capabilities of this system, we first note its fundamental difference from the previous system: the pixels in the linear array are generated simultaneously rather than serially. Thus the commutation rate

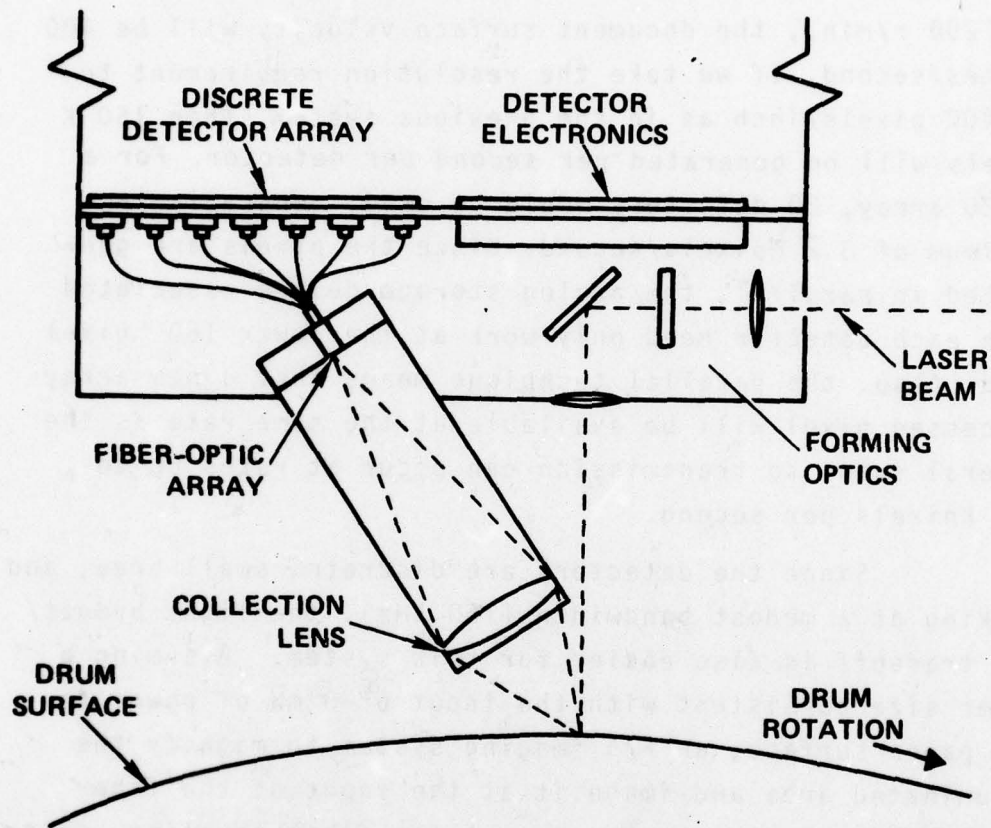


Figure 2.2.2  
Scanner Head Details for  
Drum-Scan/Fiber-Optic Array System



(bandwidth in this case) of the detectors is also the lateral pixel rate of the drum's surface. We may thus expect that this system will be capable of much higher speeds, and that the speed limiting component will be the drum itself. Assuming a drum diameter of 6.4 inches and a rotation rate of 1200 r/min., the document surface velocity will be 400 inches/second. If we take the resolution requirement to be 400 pixels/inch as in the previous system, then 160 K pixels will be generated per second per detector. For a 20x20 array, 20 detectors would be used, generating a maximum of 3.2 Mpixels/second. Since the pixels are generated in parallel, the analog storage device associated with each detector need only work at the lower 160 Kpixel rate. Also, the parallel technique means that a new array processed pixel will be available at the same rate as the lateral scan, so transmission can occur at rates up to 160 Kpixels per second.

Since the detectors are discrete, small area, and working at a modest bandwidth (160 kHz), the light budget/SNR tradeoff is also easier for this system. Assuming a laser size consistent with the input of 5 mW of power to the paper surface, an F/5 imaging system to magnify the illuminated area and image it at the input of the fiber array, and a 3 dB loss to account for fiber coupling, paper reflectivity, etc., we can expect to see approximately 0.6  $\mu$ W of optical power reaching each detector. This is in a range where at least a 35 dB SNR can be supported by many available detector/amplifier combinations, so further analysis will not be done in this cursory tradeoff consideration.

A summary of the parameters which appear to be readily supportable by an array scanning system of this type is given in Table 2.2.2.

Table 2.2.2. Parameters for Drum-Scanned  
Fiber-Optic Array System

Number of Elements	20
Array Size	20x20
Resolution (pixels/inch)	400
Detector Bandwidth (kHz)	160
Pixel Generation Rate (MHz)	3.2
Processed Pixel Transmission Rate (Kpixels/s)	$\leq 160$
Drum Speed (r/min.)	.1200
SNR (dB)	$\geq 35$

The capability to support considerably higher system data rates is the chief advantage of this system. It also relaxes the speed requirements on the detectors and analog storage devices (as compared to the previous system). The price paid for this relaxation, of course, is the need for multiple detectors and storage devices, as well as the use of a laser and a somewhat more complex optical system. For small arrays, however (up to 25X25 or so), this added complexity would not incur too steep a cost penalty; thus this scheme should be attractive for applications requiring high-speed, medium resolution array scanning.

### 2.3 GALVO-SCAN/AOBD SYSTEM - DETAILED TRADEOFFS

In the brief descriptions of the two systems of the previous section, we have seen that a drum-scanner based approach to the array sensing problem offers several advantages in the areas of design simplicity and speed. For a variety of reasons, including compatibility with previous formats and the desire for wider fields or larger matrix arrays, other system designs may be desirable. In this section we will provide a detailed tradeoff study of the component interrelationships involved in such a system. This

will provide an opportunity both to introduce the characteristic applications of some components not present in the systems of the previous section, and to exhibit in some detail the kind of tradeoff calculations which must be performed in the preliminary design of the electro-optic portions of an array scanning system. We will first adopt a set of parametric goals toward which a system can be designed, then select an approach compatible with those goals, and finally launch into the detailed tradeoff analyses needed to characterize the performance of the system.

#### 2.3.1 Adoption of Parametric Goals

Since the components and techniques that support array scanning systems have a broad range of capabilities, a more detailed design study cannot be undertaken until a set of system goals is defined with respect to which the components can be evaluated. For this purpose we have adopted the technical requirements of the Intra-Theater Imagery Transmission System (IITS) as the goals for the current array scanner. This selection was made because the IITS (although not presently in active procurement) appears to be a representative example of the current and near future imagery transmission systems needed by the image analysis community.

The goal of the IITS is a high resolution system which can achieve a minimum transmission time over limited bandwidth data links. Thus the technical parameters specified in the IITS design are relevant to both the scanning/recording portion of the Array Scanning program and the data compression and processing portion. The following table lists



the system parameters which define the required performance of the scanner system.

Table 2.3.1. IITS System Requirements (Scanner)

Scan Width (inches)	9.0
Scan Time (msec)	13.5
Digitization (bits/pixel)	4.0
Resolution (lines/inch)	384.0
Data Compression Ratio	4:1
Transmission Rate (kb/s)	32.0
MTF (@ 7.6 cycles/mm)	0.5
SNR (dB)	35.0

Additional system requirements included the ability to handle both opaque and transparent input documents, a pixel registration accuracy of  $\pm 0.10$  pixel, and a less than 5 minute warmup time.

When the focus of attention is narrowed to those systems and techniques which can support these goals, the design requirements for an array scanner become clearer. Specifically, a straightforward extension of the system proposed by Harris for the IITS, based on the use of an acousto-optic beam deflector and analog serial-to-parallel conversion, meets the array scanning requirements and offers several attractive features not available from the unmodified system. The details of a design tradeoff study of this array-scanning-modified system are given in the following paragraphs.

### 2.3.2 Approach Selection

The driving requirement in the system requirements described above is the field width of 9 inches. Except for a

few sophisticated (and relatively expensive) techniques such as the acousto-optic travelling wave lens scanners, the only scanning technique compatible with such a wide field is the galvanometer scanner. Since galvo scanning can also satisfy the scan time and resolution requirements of the system, it becomes the first component selected for the current system. To provide array sensing capability across a wide field requires either a sophisticated optical system to "descan" the reflected light so that it can be collected at a stationary array detector, or the use of a serially-scanned matrix array approach, with a single long detector and a form of storage to permit simultaneous processing of all pixels in the matrix. The latter technique offers lower risk and cost, and provides a good contrast to the schemes detailed in the previous section; it is therefore the selected approach.

### 2.3.3 Description of Selected Approach

After a design approach to be taken to produce an ITTS-oriented array scanning system had been selected, a number of specific investigations of the approach were conducted. These included preliminary tradeoff analyses to verify the feasibility of the approach, analyses of the performance of the systems key components, and brief experimental testing of several of the recommended devices which were available during the study. In this paragraph, we report on the results of these analyses, beginning with a description of the scanning system and the recommended dithered raster scanning technique.

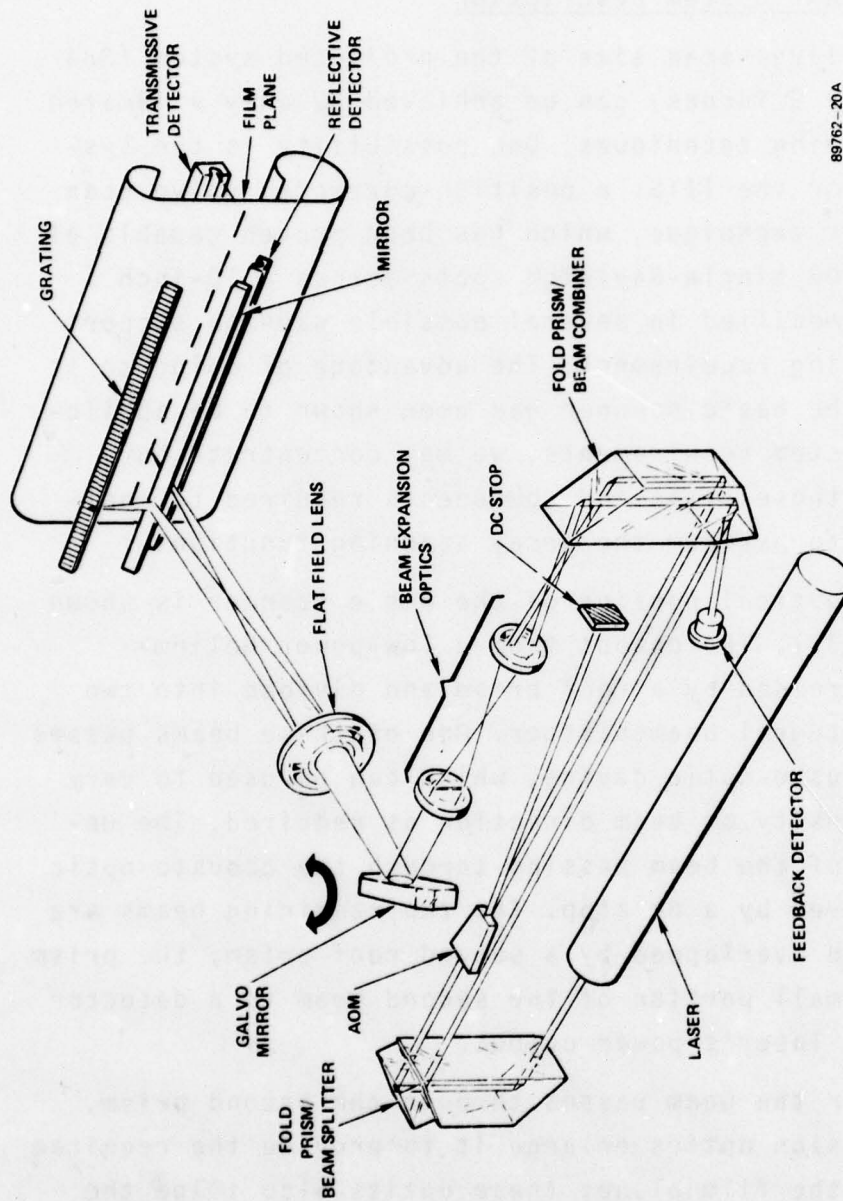
#### 2.3.3.1 Scanner System Description

The large scan size of the projected system (384 lines/inch over 9 inches) can be achieved by only a limited number of scanning techniques. One possibility is the system proposed for the IITS: a position-corrected galvo scan technique. This technique, which has been proven capable of resolving 10,000 single-Rayleigh spots across a 10-inch field, can be modified in several possible ways to support an array scanning requirement. The advantage of doing so is clear: since the basic scanner has been shown to be applicable to the system requirements, we may concentrate our evaluation on those areas and components required to modify the system to perform the array scanning functions.

The optical portion of the basic scanner is shown in Figure 2.3.3.1. The output from a low-power Helium-Neon laser is folded by a roof prism and divided into two beams by an integral beamsplitter. One of these beams passes through an acousto-optic device, which can be used to vary the light intensity or beam direction as required. The undeviated part of the beam passing through the acousto-optic device is removed by a dc stop. The two remaining beams are then folded and overlapped by a second roof prism; the prism also sends a small portion of the second beam to a detector to monitor the laser's power output.

After the beam passes through the second prism, the beam expansion optics enlarge it to provide the required resolution at the film plane; these optics also image the overlap plane (after the second prism) to the galvo mirror plane, minimizing the required mirror size, and hence, providing the fast response time required by the system.





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Figure 2.3.3.1  
Scanner System Optical Layout

Angular scanning is then imparted to the beam by the galvanometer, and the flat field lens produces a small scanned spot across the wide field with adequate depth of focus to permit easy input copy positioning in the film plane. A second function of the flat field lens is to separate the two beams in the region of the input copy plane. If the input copy is opaque, the scanning beam is reflected from it onto a long, narrow photodetector; this signal is then converted to a digital signal that carries the information relating to the imagery. (For transparencies, a second detector, located behind the film plane, would perform this function.) The second beam is reflected by a mirror and focuses onto a transparent grating which is attached to an additional long, narrow photodetector. The addition of this beam produces several important and unique performance improvements which are central to the production of a high-resolution galvanometer-based scanner.

Since the grating is configured to be in the same (equivalent) plane as the input copy, the result is two focused beams scanning simultaneously, one being used to determine the reflectance at each pixel location, the other in passing through the grating, providing information about the exact location of the pixels. Thus the grating detector provides a grating clock which is used to synchronize the data handling electronics to the received data. There are three primary design advantages to the grating clock approach. First, in spite of the nonlinear scan velocity produced by the geometry of the flat field scan, the grating clock information can be used to remove all velocity errors and simulate a completely linear scan; thus the distortion and image size tolerances can easily be met. Second, the start-of-scan

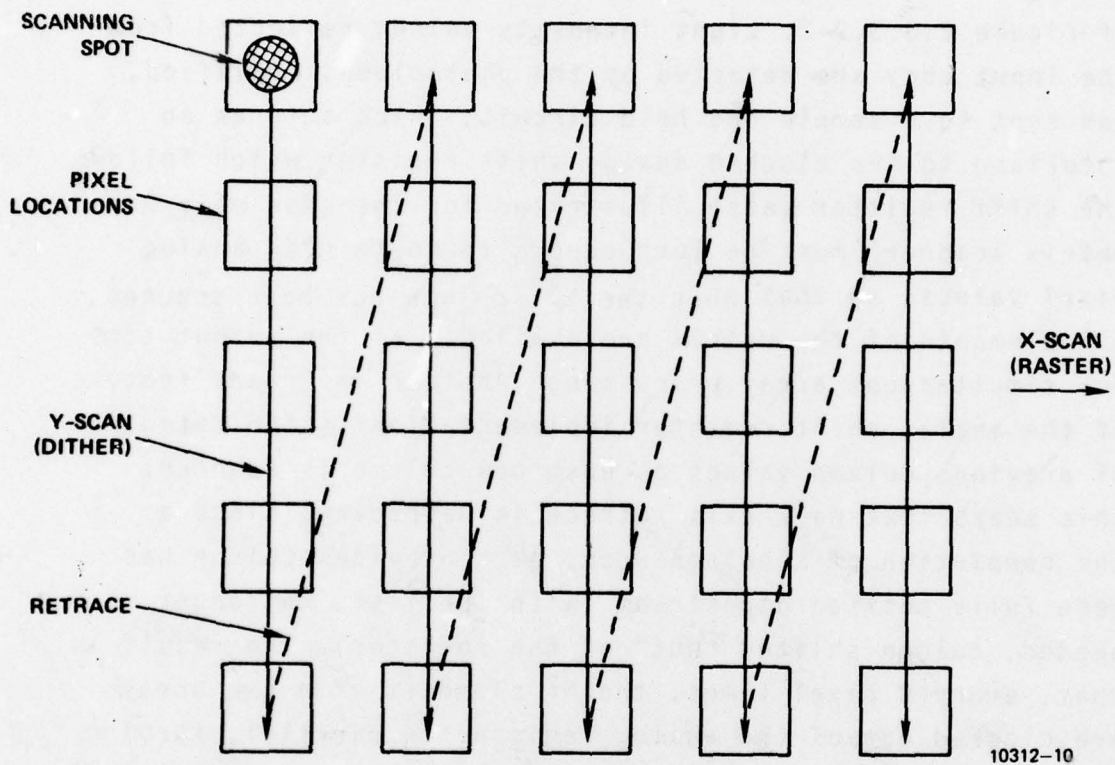
uncertainty exhibited by conventional galvo scanners ("galvo jitter") can be eliminated using the grating clock approach; thus a conventional galvo scanner can be made compatible with a high resolution system requirement. Third, the mid-scan angular velocity changes which are exhibited by some galvo scanners can also be effectively eliminated by the grating clock technique.

As discussed to this point, the scanner is essentially identical to that used in various linear-raster scanning applications. We will now turn our attention to the modifications necessary to enable it to support the array scanning requirement, while still providing the capabilities required of the IITS system.

#### 2.3.3.2 Dithered Raster Scanning

One convenient way of obtaining data for array-format processing is dithered raster scanning. This technique, which avoids both the full-line buffering required of linear raster scanners and the development of complex multichannel components needed for true matrix scanners, is based on a scanning pattern illustrated in Figure 2.3.3.2-1. The main raster scan travels in the x-direction as always, but with a vertical oscillation added. The downward strokes of the vertical motion carry the beam across  $N$  pixels (5 in the figure), after which a retrace stroke restores it to the top of the next  $N$ -element column. After  $N$  columns have been scanned in this manner, all the data associated with an  $N \times N$  array have been obtained and may be immediately processed, provided they have been converted from serial to parallel format in some sort of storage device.





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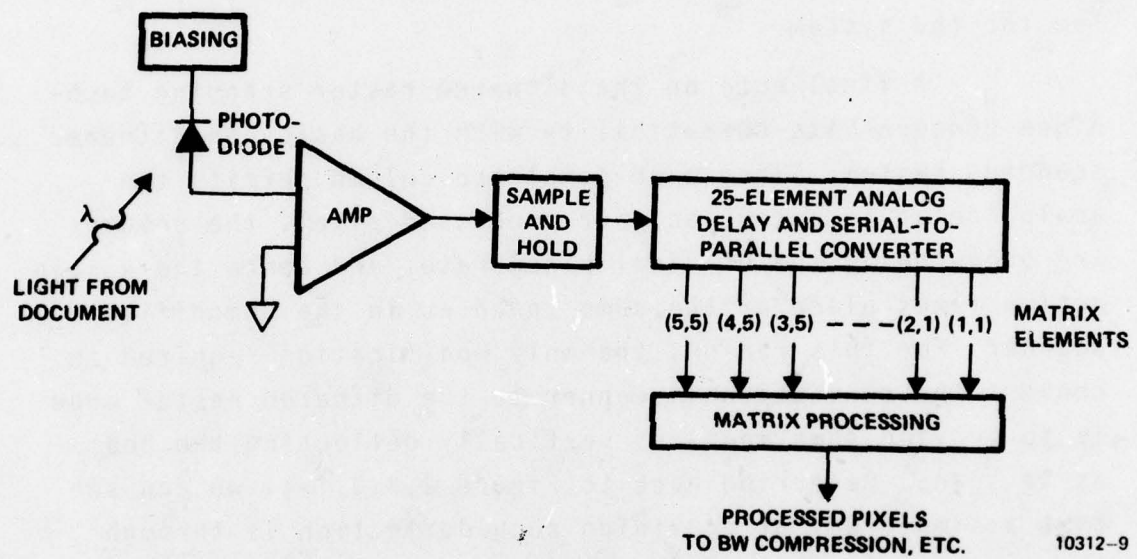
Figure 2.3.3.2-1

Dithered Raster Scanning Pattern

Thus a key to the success of the dithered raster technique in array scanning is the use of an analog shift register and serial-to-parallel converter.

The operation of a dithered raster scanner with analog pixel storage is illustrated in the block diagram of Figure 2.3.3.2-2. Light intensity values reflected from the input copy are detected by the photodiode, amplified, and sent to a sample and hold circuit, which acts as an interface to the clocked analog shift register which follows. The shift register (also illustrated for the case of a  $5 \times 5$  matrix scanner) must be long enough to contain  $N^2$  analog pixel values, so that when the  $N^{\text{th}}$  column has been scanned, all elements of the matrix are available at the output taps for simultaneous array processing. Another important feature of the analog shift register implementation is the retention of previous column values as each new column is scanned; this means that no x-axis retrace is necessary, since at the completion of a column scan, each previous column has been fully shifted downstream (with the last, no longer needed, column shifted "out" of the register). The result is that, every  $N$  pixel times, the  $N^2$  elements of a new array are clocked out of the analog register in parallel, forming a pseudomatrix indistinguishable from the output of a true matrix scanner.

The cost of achieving the simplicity and flexibility of dithered raster scanning comes in terms of the speed at which the components must operate. Since the pixel values are generated serially, the detector and other components must be capable of operation at the pixel rate, rather than the array rate as would be the case for a true array scanner. Note, however, that, because of the column storage



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Figure 2.3.3.2-2

Matrix Formatting of Scanned Information



performed by the analog shift register, its rate need be only  $N$  times the array rate (as opposed to  $N^2$  for a non-shifting storage device). As we shall see in the design tradeoff analyses to follow, the high speed components required to permit the application of the dithered raster approach to the current system goals are currently available as standard or custom devices; thus the speed requirement imposed by the technique will not be a serious problem for the system.

A final note on the dithered raster scanning technique concerns its compatibility with the basic non-dithered scanning system. Since each completed column permits the analog delay to output another processed pixel, the columns are produced at the original pixel rate, and hence the x-scan action takes place at the same speed as in the unmodified scanner. For this reason, the only modification required to convert the conventional scanner to the dithered raster mode is to provide some means of vertically deflecting the beam as it scans. Referring back to Figure 2.3.3.1-1, we can see that a simple way of providing such deflection is through the use of the acousto-optic device. If this device is configured as an acousto-optic beam deflector (AOBD), then it can provide an angular deflection of the beam in the plane between the two prisms; and since a Fourier relationship exists between that plane and the film plane, the angular deflection will produce a corresponding positional change of the scanning beam at the film. By proper orientation of the AOBD, we can insure that the deflections so produced will be vertical, producing a dithering of the raster as it sweeps horizontally across the film. Thus we have shown that a simple modification of the grating-feedback galvanometer

scanning system can produce a dithered raster system compatible with the array scanning requirement but requiring no full line buffer memory. In the paragraphs to follow, we will investigate the tradeoffs and limitations affecting the design of such a system, with emphasis on determining the maximum array size feasible with currently available components.

#### 2.3.3.3 Key Component Tradeoffs

The optical design of the system under consideration involves relatively simple components, and will not have any impact on the system's performance; thus the performance limitations of the system will be set by the active components: the laser, AOB, galvo, detector, and analog shift register. Two of these will receive only minimal consideration in the following analyses: the laser, a reliable low-power device, will only provide a light intensity value for use in the SNR and light budget calculations; also, the galvo, since its requirements are identical to those in original application (the IITS, whose goals we have adopted), will not be considered. The remaining devices will be thoroughly covered, particularly in terms of their abilities to support the necessary pixel rate, which we have seen to be the critical factor affecting the dithered raster design approach.

##### 2.3.3.3.1 AOB Tradeoffs

An acousto-optic device is a block of transparent material (various kinds of glass or crystal, for example) to which is bonded an electromechanical transducer. By driving

the transducer with an appropriate source of RF energy, travelling pressure waves can be created within the crystal which are capable of deflecting a light beam. This occurs because the changes in pressure in the crystal are accompanied by changes in the index of refraction, making the crystal appear to the light beam as a phase-only diffraction grating. The light diffracted from this grating can be rapidly controlled by controlling the RF input: amplitude variations can change the beam's intensity, while frequency variations can change its angle of diffraction. The latter effect is the basis of operation of an acousto-optic beam deflector (AOBD).

The design of the AOBD for the dithered raster scanner will affect the achievable performance in several key areas. The risetime of the deflector, because it represents the minimum time between pixel locations for the scanning beam, will set a limit on the system's speed. The diffraction efficiency of the device will impact the amount of input light required to provide a given SNR at the detector, and hence the system light budget and detector bandwidth trade-offs will be affected. Finally, the bandwidth of the AOBD itself, because it controls the number of resolvable spots the device can produce in a given time, will directly place an upper limit on the size of the matrix array achievable in the system.

The optical system required to produce angular deflection in the collimated beam between the two prisms is shown in Figure 2.3.3.3.1. The beam is assumed to be initially 1 mm in diameter, and is demagnified by the two input lenses to minimize the risetime in the AOBD. After



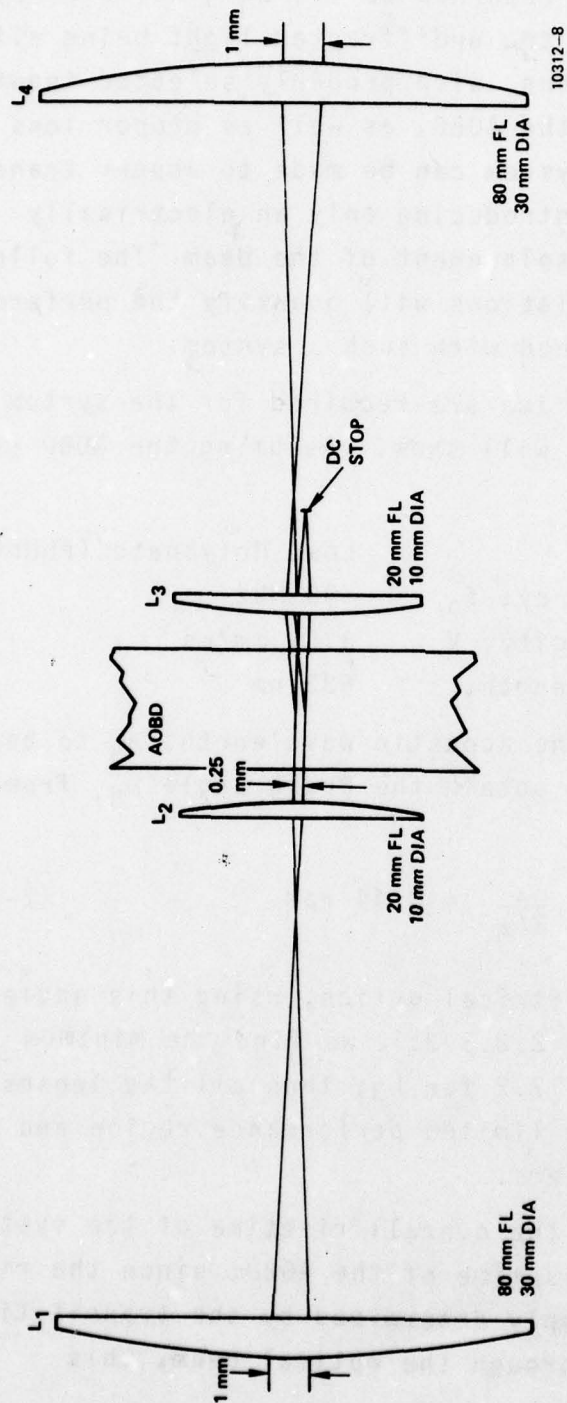


Figure 2.3.3.3.1  
AOBD Optical Design

deflection, the beam is returned to its original size by the output lenses, with the undiffracted light being eliminated at a dc stop. Thus, with properly selected input and output surfaces on the AOB, as well as proper lens positioning, the AOB system can be made to appear transparent to the system, introducing only an electrically controllable angular displacement of the beam. The following simple design calculations will quantify the performance which may be obtained with such a system.

Optics. No critical optics are required for the system, as a simple calculation will show. Assuming the AOB parameters are as follows

Material	Lead Molybdate ( $\text{PbMoO}_3$ )
Center Frequency, $f_0$	500 MHz
Acoustic Velocity, $V$	3.61 mm/ $\mu$ s
Optical Wavelength, $\lambda$	633 nm

then we can calculate the acoustic wavelength,  $\Lambda_A$  to be about 7.2  $\mu$ m, and thus obtain the Bragg angle  $\theta_B$ , from the approximation

$$\theta_B \approx \sin \theta_B = \frac{\lambda}{2\Lambda_A} = 0.44 \text{ rad.} \quad (2-2)$$

By straightforward geometrical optics, using this angle in the system of Figure 2.3.3.3.1, we find the minimum F# in the system to be 17.2 for  $L_3$ ; thus all the lenses are used in the diffraction limited performance region and pose no optical design problems.

Risetime. To minimize the overall risetime of the system, we must minimize the risetime of the AOB; since the risetime of the AOB is simply determined by the transit time of the acoustic wave through the optical beam, this

means a narrow optical beam is required. However, to provide a positional deflection at the film plane, the AOBD must be located in a collimated beam, so focusing through the AOBD to minimize risetime is not an available option. As illustrated in the figure, the beam may be demagnified by about a factor of four using two lenses; for greater demagnifications, the diffraction spreading of the small beam would re-expand the beam before it reached the acousto-optic interaction region, defeating the attempt to achieve a smaller risetime in a collimated beam. Assuming a four-to-one reduction is used as illustrated, and a 20 mm distance from the second lens to the center of the crystal, the optical beam interacting with the acoustic wave will be approximately 258  $\mu\text{m}$  in diameter. Therefore the minimum risetime achievable will be given by

$$T_r = \frac{258 \mu\text{m}}{3.61 \mu\text{m/ns}} \cong 71 \text{ ns.} \quad (2-3)$$

Time-Bandwidth Product. An additional consideration in the design of the AOBD is the bandwidth of the device. For operation in the Bragg mode (for high efficiency), the acoustic "Q" of the device should be at least  $4\pi$ . Here Q is defined as

$$Q = \frac{2\pi\lambda L f_0^2}{nV^2}, \quad (2-4)$$

where L is the transducer length and n is the index of refraction. For lead molybdate and the previously specified parameters, setting Q equal to  $4\pi$  gives

$$L f_0^2 = 101.3 \mu\text{m}(\text{GHz})^2. \quad (2-5)$$



Then if we select a center frequency,  $f_0$ , of 500 MHz, which is near the upper limit of current standard acousto-optic practice, we get the required transducer length,  $L$ , to be 405 mm. Finally, using the relation

$$\Delta f_b = \frac{1.8 \cos \theta_0 (2\pi f_0)}{Q} \quad (2-6)$$

We can find the 3 dB bandwidth of the device to be given by  $\Delta f_B = 450$  MHz. Since the number of spots resolved by the AOBBD will be equal to the time-bandwidth product, i.e.,  $N \leq T_r \Delta f_B$ , we arrive at the constraint

$$N \leq (0.45) T_r, \quad (2-7)$$

where  $T_r$  has been expressed in nanoseconds.

#### Diffraction Efficiency

For later use in the light budget calculations, we now determine the efficiency of the projected AOBBD using the equation

$$\eta = \sin^2 \frac{\pi}{\lambda} \frac{M_2 L}{2H}, \quad (2-8)$$

where  $M_2$  is the figure of merit of the acoustic material ( $36.1 \times 10^{15} \text{ m}^2/\text{W}$  for lead molybdate),  $H$  is the transducer height, and 1 watt of acoustic input power has been assumed. Taking  $H = 250 \text{ } \mu\text{m}$  to match it to the size of the optical beam, we get

$$\eta = 56\% \quad (2-9)$$

## Summary

The brief analyses presented above have shown that an AOBBD can be designed which can support the array scanning/dithered raster system with risetimes as low as 71 ns and efficiencies as high as 56% per watt. Having determined that the projected design is feasible, we now need only check the compatibility of the achieved parameter values with the other constraint analyses (risetimes, light budget, etc) to follow.

### 2.3.3.3.2 Detector Tradeoffs

Another key component in the dithered raster scanner is the long scan detector which receives the light reflected from the surface of the input document. Currently, only one device, the UDT-Pin-L9 detector, is capable of supporting both the scan length (9 inches) and speed requirements of the system being analyzed. Because the system's feasibility depends on the performance of this device, detailed analyses of its bandwidth and sensitivity were performed, as well as some preliminary experiments to verify the detector's performance at high speeds. The following paragraphs summarize the results of these investigations.

### Bandwidth-Noise Relationship

The nominal specifications of the PIN-L9 detector are given by the manufacturer as

NOISE EQUIVALENT POWER (nW/ MHz)	50
BANDWIDTH (MHz)	2
INPUT CAPACITANCE, 50V BIAS (pF)	900
RESPONSIVITY (A/W)	0.3

To determine the performance to be expected of this device in practice, we must analyze it in combination with an appropriate preamplifier circuit. The model circuit chosen for analysis is shown in Figure 2.3.3.3.2-1. The detector input capacitance,  $C_i$ , has been included, as well as the transfer impedance,  $R$ , and the current and voltage noise figures of the amplifier. In this situation the primary noise sources and their values are:

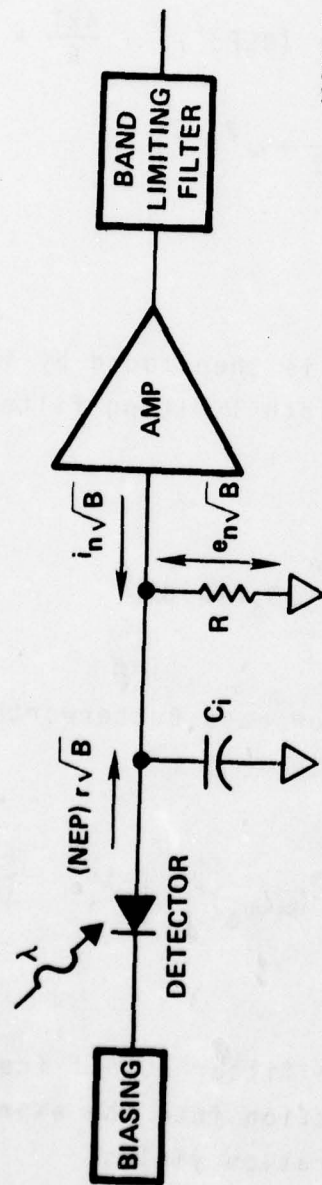
<u>NOISE SOURCE</u>	<u>EXPRESSION</u>
Photodiode Noise Current	$(NEP) \times r \sqrt{B}$
Resistor Thermal Noise Current	$\sqrt{4kTB/R}$
Amplifier Noise Current	$i_n \sqrt{B}$
Amplifier Noise Voltage	$e_n \sqrt{B}$

Here,  $B$  is the bandwidth as limited by the filter,  $k$  is Boltzmann's constant, and  $T$  is the temperature in  $^{\circ}K$ . According to this model, the photodiode detector is seen as a current source; its impedance is a large value of resistance shunted by a capacitor. The system bandwidth is determined by the associated circuitry.

The signal-to-noise ratio may be determined by finding the equivalent noise input current of the system and comparing it with the signal current. Simple circuit analysis of the circuit in Figure 2.3.3.3.2-1 shows that the input noise current equivalent to the noise voltage of the amplifier is given by

$$i_{eq}^2 = e_n^2 \left( \frac{1}{R_i^2} + \omega^2 C_i^2 \right), \quad (2-10)$$





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Figure 2.3.3.3.2-1  
Model Circuit for Detector Analysis

where  $\omega$  is the radian frequency and  $R_i$  is the total input resistance. Then the total input noise current spectral density will be given by:

$$S_i = \frac{d(i_i^2)}{d\omega} = (\text{NEP})^2 r^2 + \frac{4kT}{R} + i_n^2 + e_n^2 \left( \frac{1}{R_i} + \omega^2 C_i^2 \right). \quad (2-11)$$

The total input noise current is then found by integrating over frequency with the bandwidth limiting filter function as a weighting factor, i.e.,

$$i_i^2 = \frac{1}{2\pi} \int_0^\infty S_i F(\omega) d\omega. \quad (2-12)$$

If the filter has an n-pole low pass Butterworth characteristic, then

$$F(\omega) = \left[ 1 + (\omega/\omega_0)^{2n} \right]^{-1}, \quad (2-13)$$

where  $\omega_0 = 2\pi f_0$  and  $f_0$  is the filter cutoff frequency. Substituting this filter function into the expression for  $i_i^2$  and performing the integration yields

$$i_i^2 = \left[ (NEP)^2 r^2 + \frac{4kT}{R} + i_n^2 + \frac{e_n^2}{R_i^2} + K_1 (2\pi B_n C_i e_n)^2 \right] B_n \quad (2-14)$$

where we have defined

$$B_n \triangleq f_o / \text{sinc}(1/2n) \quad \text{and} \quad K_1 \triangleq [\text{sinc}^3(1/2n)] / 3\text{sinc}(3/2n)$$

and

$$\text{sinc}(x) \triangleq \frac{\sin(\pi x)}{(\pi x)} .$$

This final equation for  $e_i^2$  relates the input noise current to the noise sources and the noise bandwidth.

To make this result useful for our tradeoff study, we will now assume the use of a specific device, so that the expression for  $i_i$  may be simplified. For a low-noise J-FET device of type LF-157, the following conditions hold:  $e_n = 15 \times 10^{-9}$  V;  $i_n = 1 \times 10^{-14}$  A; high input impedance implies that  $R_i = R$ . For the detector under study, we also have  $C_i = 900$  pF,  $NEP = 50$  nW/Hz,  $r = 0.3$  A/W; we will further assume that  $R = 1.1$  k $\Omega$ . Inserting these values into the expression for  $i_i$  yields

$$i_i \simeq (2.06 \times 10^{-11}) \sqrt{B_n} \sqrt{1 + (1.69 \times 10^{-11}) K_1 B_n^2} . \quad (2-15)$$

In this expression, the  $K_1 B_n^2$  term becomes significant as  $B_n$  approaches 300 kHz; since we are interested in the high frequency performance limits, we may further simplify to obtain



$$i_i \approx (8.4 \times 10^{-17}) \sqrt{K_1} B_n^{3/2}. \quad (2-16)$$

Finally, if we assume the use of a 2-pole Butterworth filter, and express  $i_i$  in microamperes and  $B_n$  in Megahertz, we obtain

$$i_i \approx (0.076) B^{3/2}. \quad (2-17)$$

This result will be of use in a subsequent determination of the system SNR and light budget.

#### Detector Collection Efficiency

The light reflected from the input document propagates in all directions, and can therefore be only partially collected by the scan detector. The amount that can be detected is an important consideration for the system efficiency calculations to come. Therefore we will briefly consider the geometrical restrictions imposed by the detector size and limitations. Figure 2.3.3.3.2-2 shows a schematic of the PIN-L9 detector unit; the dimensions are specified by the manufacturer as  $L=9.0$  inches,  $W=0.5$  inches, and  $d=0.1$  inch. Figure 2.3.3.3.2-3 is a model of the situation which will be used for calculation. To simplify the calculation, the model assumes that the detector lies on the surface of a cylinder which surrounds the scan line. From the surface of the input document, the reflected light goes into  $2\pi$  steradians. Furthermore, for the worst case (as illustrated)

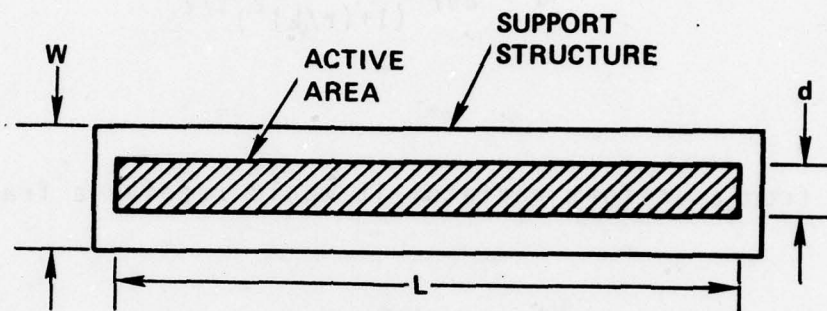
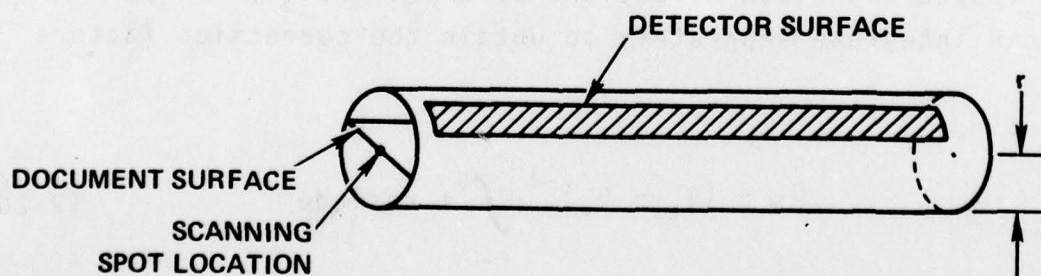


Figure 2.3.3.3.2-2  
Scan Detector Geometry



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Figure 2.3.3.3.2-3  
Model for Collection Efficiency Calculation

where the scanning spot is at the end of the detector, the angular subtense of the detector as seen at the spot is

$$\Omega_d \approx \frac{d}{2\pi r} \frac{2\pi}{(1+(r/L)^2)^{1/2}} \quad (2-18)$$

so from a uniform radiator it would receive a fraction

$$\frac{\Omega_d}{2\pi} \approx \frac{d/2\pi r}{(1+(r/L)^2)^{1/2}} \quad (2-19)$$

of the light. Since the reflected light will not be uniform, but will approximately obey Lambert's Law (i.e., proportionality to the cosine of the angle from the scattered direction to the surface normal), we must add another correction factor to account for the detector's average direction. By treating the length and width directions as approximately independent, we can integrate separately to obtain the correction factors

$$C_W = (\theta_+ - \theta_-)^{-1} \int_{\theta_-}^{\theta_+} \cos \theta d\theta \quad (2-20)$$

$$C_L = (\theta_L)^{-1} \int_0^{\theta_L} \cos \theta d\theta, \quad (2-21)$$



where  $\theta_{\pm} \triangleq \tan^{-1} (W \pm d/2r)$  and  $\theta_L \triangleq \tan^{-1} (L/r)$ . Thus the total collection efficiency is approximately

$$\eta_d \approx \frac{1}{2\pi} \Omega_d C_W C_L. \quad (2-22)$$

For the PIN-L9 (UDT) detector,  $W=0.5"$ ,  $L=9"$ ,  $d=0.1"$ , and we will assume  $r=0.25"$ . This gives

$$\eta_d \approx 6.1\%, \quad (2-23)$$

or with a 90% reflective naper,

$$\eta_{TOTAL} \approx 5.5\%. \quad (2-24)$$

A final consideration is derived from experience with facsimile systems, where it has been found that the use of a mirror at each end of the scan can approximately double the collected light. Accordingly we will adopt for further calculation a final detector collection efficiency value of

$$\eta \approx 10\%. \quad (2-25)$$

#### Detector Experiments

Having established that the PIN-L9 detector could theoretically support reasonably high rates of scanning, we next tested an available unit for verification of these results. The circuit shown in Figure 2.3.3.3.2-4 was used

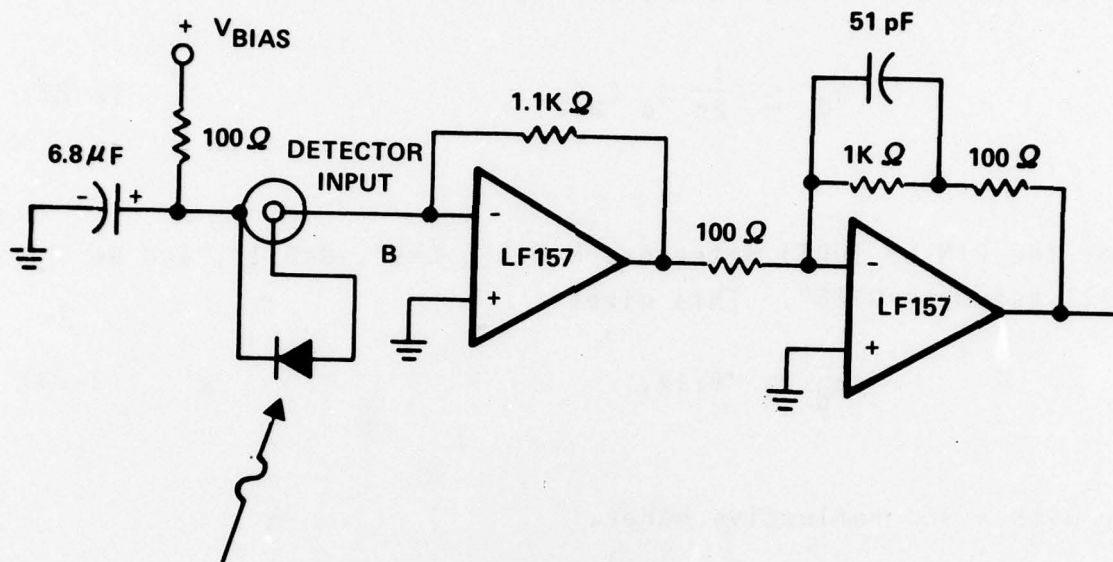


Figure 2.3.3.3.2-4

Test Circuit for PIN-L9 Detector

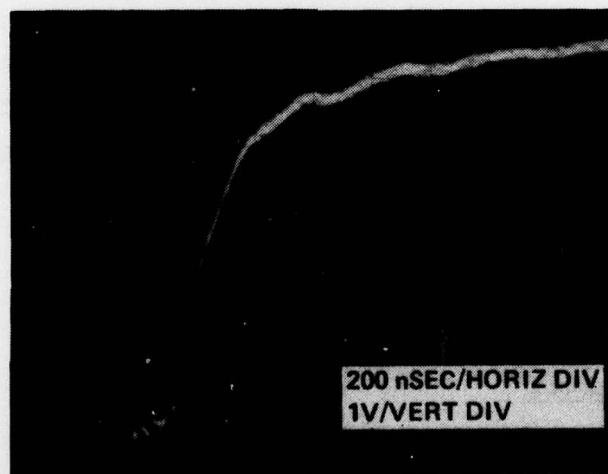


Figure 2.3.3.3.2-5

Detector Step Response

for the test; the detector was used in conjunction with a pair of LF-157 amplifiers. For input, the detector was illuminated with a laser beam which was modulated by an acousto-optic modulator. In this way the risetime of the detector could be measured and its corresponding frequency response calculated. The step response measured in this way is shown in Figure 2.3.3.3.2-5. By measuring the amplitude values shown for various time locations, a set of data was collected for input to a Fourier transform program to determine the frequency response. The result of this computation is shown in Figure 2.3.3.3.2-6. The 3 dB bandwidth of the detector is seen to be only about 700 KHz; however, the detector response remains monotonic until about 1.1 MHz, at which point its signal is down by about 7 dB. Thus there is hope that a frequency-compensating filter network can be constructed to flatten the detector response out to slightly above 1 MHz. This experiment does indicate, however, that the detector's specification (2 MHz) appears to be optimistic, and that a lower value should be used for any critical system design.

#### 2.3.3.3.3 TAD Performance

Because it is a central element in the scanning system, a brief examination of the performance of a currently available tapped analog delay (TAD) device was made; this paragraph will document both specifications and conceptual operation of the device, as well as the results of the experiments conducted with it.

The device to be tested was the TAD-32, a product of the Reticon Corporation, along with the associated circuitry of the TC-32 Evaluation Circuit Card. The TAD-32 is described by Reticon as follows:



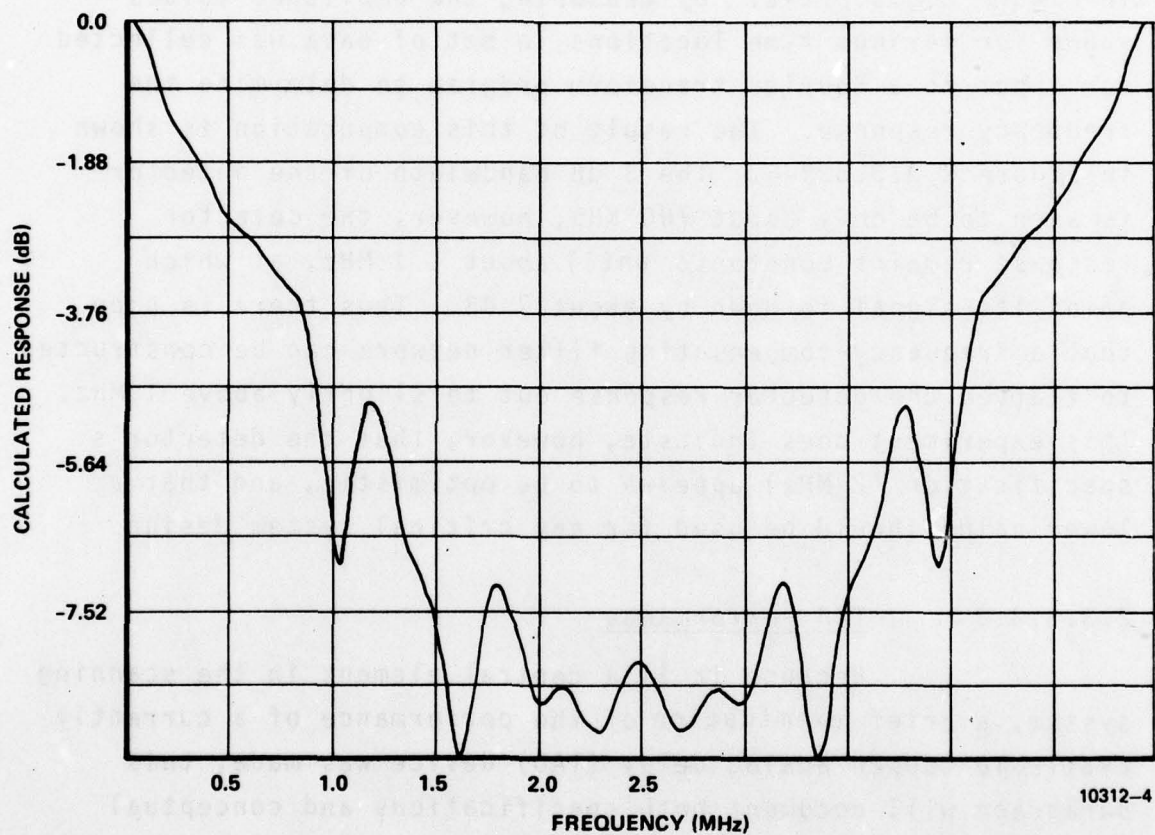


Figure 2.3.3.3.2-6  
Detector Frequency Response Calculated  
From Step Response

TAD-32 is a 32-stage charge-transfer device which permits the storage of analog signals with recovery of the signals at multiple separate outputs at successive delay times later. The taps on each stage are brought to the outside through buffer amplifiers. Each buffer amplifier output appears as a source follower, thus permitting variable loading of the taps in order to create various tap-weight functions. The taps are spaced one sample time apart along the delay. An additional special feed-forward output tap is provided so that multiple devices may be cascaded without causing discontinuity in the spacing of the taps from one device to the next. With this arrangement, timing integrity is maintained. The ability to cascade devices permits the user to build processors (such as transversal filters) with more than 32 taps.

The device specifications (from the manufacturer's 1976 documentation) include

No. of Taps	32
Clock Frequency Range	1 KHz - 5 MHz
Max. Power Requirement	700 mW
Dynamic Range	60 dB
Sensitivity (for SNR $\approx 1$ )	4 mV p-p

Thus the TAD device appears ideally suited to the serial-in, parallel-out shifting and weighting of analog pixel values as required for the dithered raster scanner.

Since a TC-32 Evaluation Circuit Card (which includes all necessary support circuitry, tap weighting

resistors and switches, etc.) was available within Harris, some brief tests were conducted to evaluate its applicability to the array scanning requirement. The first test concerned the key parameter of clocking speed. A video signal from a television camera was input to the TAD-32 and its output was fed to a standard monitor. The tap weights were adjusted to represent various types of transversal filters (bandpass, lowpass, etc.) and the clock rate was varied over a wide range. The result showed good video fidelity at clock rates up to 10-15 MHz, well above the 5 MHz specification; it should be noted, however, that the frequency response required for visually acceptable video information is considerably less than that needed in the array scanning system; thus the accuracy of the analog samples being shifted by the device at high clock rates was not completely determined by this test. Nevertheless, since clock rates at or below the 5 MHz specification would be required for a IITS-type array scanner, this result is encouraging.

A second test was made of the pulse fidelity transmitted by the device. For this test, a single pulse was repetitively input to the device and examined as it passed each output tap. During this test, considerable pulse dispersion was observed: at the 32nd tap, the pulse has lost up to 20% of its initial voltage, while the pulses leading and trailing the main pulse had grown to about 20% of the initial pulse height (with the leading pulse voltage being negative). Discussions with Reticon Corporation after this test was completed confirmed the problem, and disclosed that a technique of compensating for it has been published by J. J. Brady of the Applied Research Lab at the University of Texas at Austin. In that publication [1](ARL-TP-77-13), it was shown



that the use of discrete transistors to provide a low impedance to ground at each tap could greatly reduce the pulse dispersion effect. An additional consideration is that the pulse dispersion effect may be of minor importance in the array scanning application. A computer simulation (documented elsewhere in this report) has shown this to be the case for a realistic choice of parameters and applications.

The performance of the TAD-32 device appears capable of supporting the array scanning application in many important cases. An additional important fact is that, for non-adjustable filter applications, analog delay lines/transversal filters have been constructed (by Reticon and others) with as many as 512 stages, indicating that pulse dispersion and other problems can indeed be made negligible in custom devices. In summary, the TAD appears to be a viable component for inclusion in the dithered raster scanning system described above.

#### 2.3.3.3.4 Light Budget and SNR

With estimates of the efficiencies of the various active components in the system now complete, we are in a position to evaluate the ability of the system to produce an acceptable SNR for a given amount of input light. Since the scanning system is identical to that of the IITS system (except for the replacement of the AOM by the AOB), we will adopt and modify the original light budget calculation, reproduced in Table 2.3.3.3.4. Replacing the AOM efficiency (65%) with the AOB efficiency,  $\eta$ , and assuming a 5 mW laser source, we find approximately that the light expected at the detector,  $I$ , is given by

$$I = \left( \begin{array}{c} \text{detector} \\ \text{collection} \\ \text{efficiency} \end{array} \right) \times (\eta) \times (2.5 \text{ milliwatts}) \quad (2-26)$$

Table 2.3.3.3.4. Detailed Light Budget

<u>Exposing Beam</u>		<u>Grating Beam</u>	
<u>Optical Element</u>	<u>Efficiency Percent</u>	<u>Optical Element</u>	<u>Efficiency Percent</u>
BS Prism Surface 1	0.998	BS Prism Surface 1	0.998
BS Prism Surface 2	0.85	BS Prism Surface 2	0.15
BS Prism Surface 3	0.998	BS Prism Surface 3	0.998
AOM (includes reflection losses)	0.65		
Folding Prism Surface 1	0.96	Folding Prism Surface 1	0.998
Folding Prism Surface 2	0.998	Folding Prism Surface 2	0.998
Telescope: Lens 1	0.98	Telescope: Lens 1	0.98
Telescope: Lens 2	0.98	Telescope: Lens 2	0.98
Galvo Mirror	0.90	Galvo Mirror	0.90
Flat Field Lens	0.96	Flat Field Lens	0.96
<u>Transmission and Truncation</u>	0.80	Fold Mirror	0.99
Total	0.35	<u>Transmission and Truncation</u>	0.80
Available exposing power (5 mW Laser)	1.76 mW	Total	0.10
		Available grating beam power (5 mW Laser)	0.50 mW

Modulation Feedback Detector

	<u>Efficiency Percent</u>
BS Prism Surface 1	0.998
BS Prism Surface 2	0.85
BS Prism Surface 3	0.998
AOM (includes reflection loss)	0.65
Folding Prism Surface 1	0.04
<u>Transmission and Truncation</u>	0.80
Total	0.02
Available Power (5 mW Laser)	0.09 mW

so with a collection efficiency of 10%, we get

$$I = 250\eta \text{ } \mu\text{W.} \quad (2-27)$$

From the previous AOB calculations we have

$$\eta = \sin^2 \left\{ \frac{\pi}{\lambda} \sqrt{\frac{M_2 L}{2H}} \right\}; \quad (2-28)$$

thus using the values  $\lambda=633 \text{ nm}$ ,  $V=3.61 \text{ mm}/\mu\text{s}$ ,  $L=0.405 \text{ mm}$ ,  $M_2=36.1 \times 10^{15} \text{ m}^2/\text{W}$ ,  $H=VT_r$ , we get

$$\eta = \sin^2 \left\{ \frac{7.0197}{\sqrt{T_r}} \right\}. \quad (2-29)$$

Thus, for a silicon detector with a responsivity of  $0.3 \text{ A/W}$ , the expected signal current is

$$i_s = (0.3)I = 75 \sin^2 \left( \frac{7.0197}{\sqrt{T_r}} \right) \mu\text{A.} \quad (2-30)$$

As shown in the detector analysis above, for a PIN-L9 photodetector in conjunction with an LF-157 amplifier and a 2-pole Butterworth filter, the expected noise current for wideband operation is given by

$$i_n \approx (B^{3/2})/13.2,$$



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ARRAY SCANNING FOR IMAGE TRANSMISSION.(U)  
JAN 79 W C ADAMS, T R BADER, J J KNAB

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where B is the bandwidth of the external filter. Therefore the current SNR is

$$\text{SNR} = i_s/i_n = (B^{-3/2})990 \sin^2(7.0197/\sqrt{T_r}). \quad (2-32)$$

The (arbitrarily selected) SNR requirement for the system is 35 dB, which implies a current ratio of 56.2. If we increase this to 60 to allow some margin, we can find the bandwidth/risetime relationship from the above equation, i.e.,

$$B = \left| 16.5 \sin^2(7.0197/\sqrt{T_r}) \right|^{2/3} \text{ MHz}. \quad (2-33)$$

Thus, finally, the risetime of the detector circuit can be related to the AOBBD risetime using

$$T_d = 350 \left| 16.5 \sin^2(7.0197/\sqrt{T_r}) \right|^{-2/3} \text{ ns}. \quad (2-34)$$

This value can now be used in the overall system risetime calculation.

#### 2.3.3.3.5 Risetimes

Within each pixel time,  $T_p$ , must be included the risetimes of all the active components; this includes the AOBBD risetime,  $T_r$ , the detector risetime,  $T_d$ , and the sample/hold risetime  $T_{s/h}$ . We will assume that these risetimes can be combined in a root-sum-square way, via the inequality

$$T_p^2 \geq T_r^2 + T_d^2 + T_{s/h}^2. \quad (2-35)$$

The pixel time,  $T_p$ , is derived from the IITS system requirements as follows. Within one 13.5 ms scan time, 3,456 pixel columns must be completed. For an  $N \times N$  matrix array scanner, each column will contain  $N$  pixels. Thus the pixel time is

$$T_p = \frac{13.5 \text{ ms}}{3456 N} = \frac{3.906 \mu\text{s}}{N}. \quad (2-36)$$

Now, if we substitute for  $T_p$ ,  $T_d$ , and  $T_{s/h}$  in the above inequality, we get the constraint

$$N \leq 3906 \left\{ (35)^2 + T_r^2 + (350)^2 \left[ 16.5 \sin^2 \left( \frac{7.0197}{\sqrt{T_r}} \right) \right]^{-4/3} \right\}^{-1/2} \quad (2-37)$$

where we have used for  $T_{s/h}$  the value of 35 ns, which is typical of currently available devices. It should be noted that this constraint inequality was derived under the assumption that the maximum theoretical bandwidth, as given in the previous sections, could be used. We have seen experimentally, however, that compensation of the detector under consideration for use at very high bandwidths (above 1 MHz) will be a difficult task. Therefore, it is also useful to state the risetime constraint as it would appear if we achieve not the maximum bandwidth, but some lower value,  $B$  (MHz). In that case, the constraint is simply



$$N \leq 3906 \left[ (35)^2 + T_r^2 + (350/B)^2 \right]^{-1/2}. \quad (2-38)$$

It should be noted also that  $T_r$  as used in these inequalities is the AOBBD transit time only when a random-access type scanning is being done; however, this is appropriate to the current scanner provided that, to simplify the analog sample delay line process, we require that all pixels be generated at regular time intervals (as opposed to burst-and-wait type sampling).

The constraint of equation (2-38) can now be used to define the sizes of matrix arrays which the dithered raster technique as described above can support. This will be done by way of the constraint summary graph presented in the following section.

#### 2.3.3.4 Constraint Summary

Four basic constraints on the feasible size of the matrix array in the dithered raster scanner can be defined. The first and simplest, the matrix constraint, can be expressed as  $N \geq 3$ , and results from the desire to have an odd-edged square array to preserve the uniqueness of the central point. Second is the optics constraint: to produce beam deflection at the film, the AOBBD must be in a Fourier plane of the scanning spot, and thus its input beam cannot be focused to minimize risetime; with 20 mm between AOBBD and input lens, the minimum beam diameter (due to Gaussian beam diffraction) is 180  $\mu\text{m}$  (50 ns risetime); restriction to 71 ns keeps the required optical F#'s reasonable, thus  $T_r \geq 71$  ns. Third is the AOBBD design constraint, which arises from the time-bandwidth product of the device; with a state-of-the-art bandwidth of 450 MHz, this produces the relation  $N \leq 0.45 T_r$ .

Finally, the detector speed and SNR constraint as derived above requires  $N \leq 3906 \left[ (35)^2 + T_r^2 + (350/B)^2 \right]^{-1/2}$ . With plenty of optical power available to achieve the required SNR, the goal is to trade it for detector bandwidth (and hence, risetime). The result is a coupling of system risetime, detector bandwidth and SNR.

The constraints above have been derived in terms of the AOBD risetime,  $T_r$ , so that it may act as a common parameter against which the array size may be plotted. The result of this is shown in Figure 2.3.3.4. The upper curved line is the risetime constraint obtained when it is assumed that the detector response can be equalized out to the maximum bandwidth for which the SNR is adequate. The lower curves are based on constant bandwidth values lower than the maximum (and hence these curves describe increasing SNR as  $T_r$  increases). Because of the experimental performance of the PIN-L9 detector, the maximum bandwidth curve is not likely. As a more practical upper bound, a bandwidth of 1 MHz seems readily achievable, and 1.6 MHz should be attainable with proper detector selection and compensation. From this we may conclude that array sizes up to  $11 \times 11$  are definitely feasible and that arrays up to  $17 \times 17$  are conceivable with careful design. Thus, the dithered raster system with analog shift register can support, at adequate speed, SNR, and resolution levels, all but the largest of array scanning requirements, and this can be done with only relatively minor modification to the proven scanning system concept as exemplified in the (proposed) IITS system.

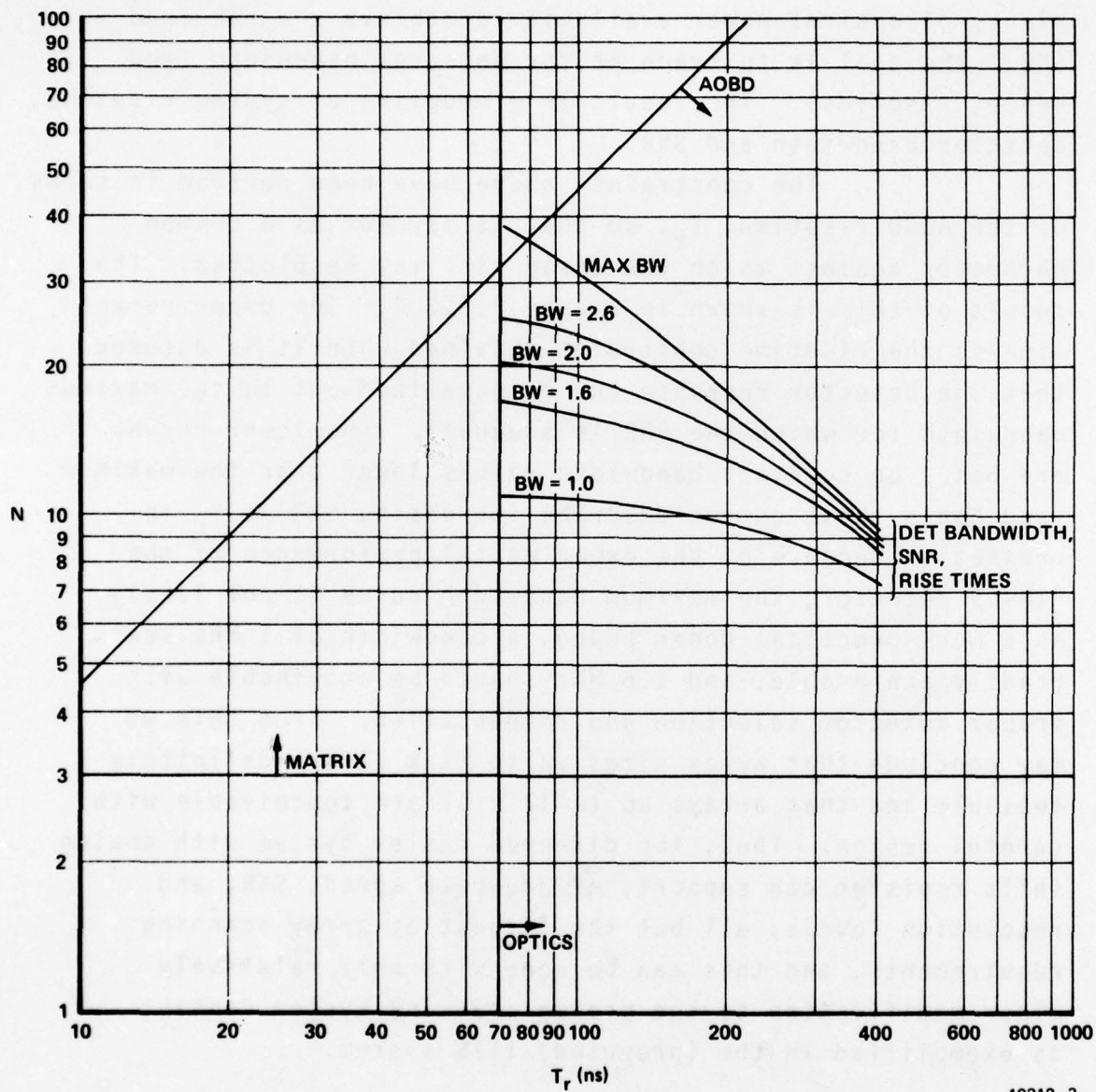


Figure 2.3.3.4  
Array Size Constraint Summary



## 2.4 ALTERNATIVE APPROACHES

During the investigation of the dithered raster array scanner described above, several additional approaches to the specific problem addressed in section 2.3 were identified which show promise for future study. Although not examined thoroughly to this point, these approaches appear to have some advantages (and disadvantages) with respect to the main system concept which would make them particularly useful in some system scenarios. We will now briefly document two of these approaches.

### 2.4.1 Weighted Linear Array with Column Memory

In the main system of Paragraph 2.3, the scanning was done by a single, dithered beam, with the pixel values of each column of the array being stored in the tapped analog delay line for later parallel processing. The price paid for the design simplicity of that system was the high speed required to generate serially the  $N^2$  pixels of the array. A more complex, but potentially much faster, technique involves the scanning of multiple beams in a linear array. If true array scanning is employed, both the input array and the detector array (required to separately detect each channel) are complex devices; however, an interesting alternative is made possible by the use of the tapped analog delay: linear array scanning with a single detector. In this case (see Figure 2.4.1-1), a linear array of spots from a multi-element source, such as an acousto-optic page composer, is scanned across the page. The scattered light from all the spots is collected into a single detector, which produces a summation process. If the pixel weights

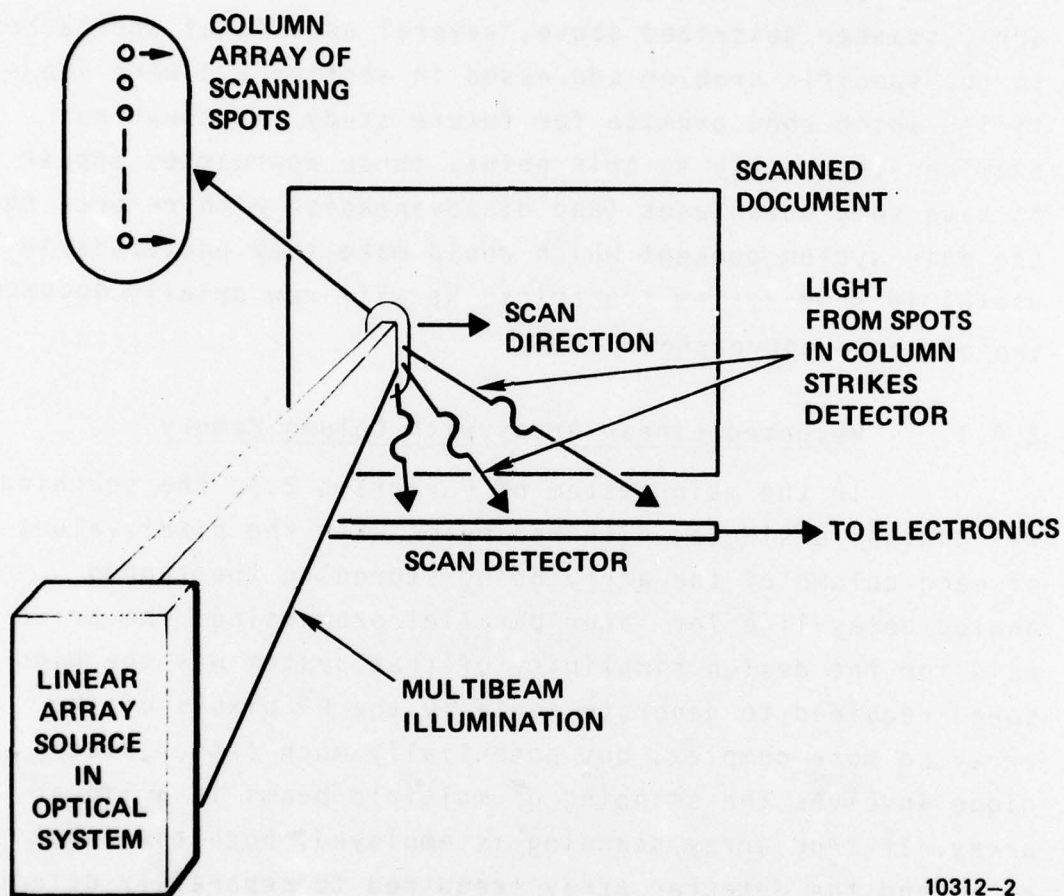


Figure 2.4.1-1  
Array Scanning With Single Scan Detector

(document reflectances) are called the  $a_{ij}$ , then the detector would generate for each column the sum

$$S_i = \sum_j a_{ij}c_j, \quad (2-39)$$

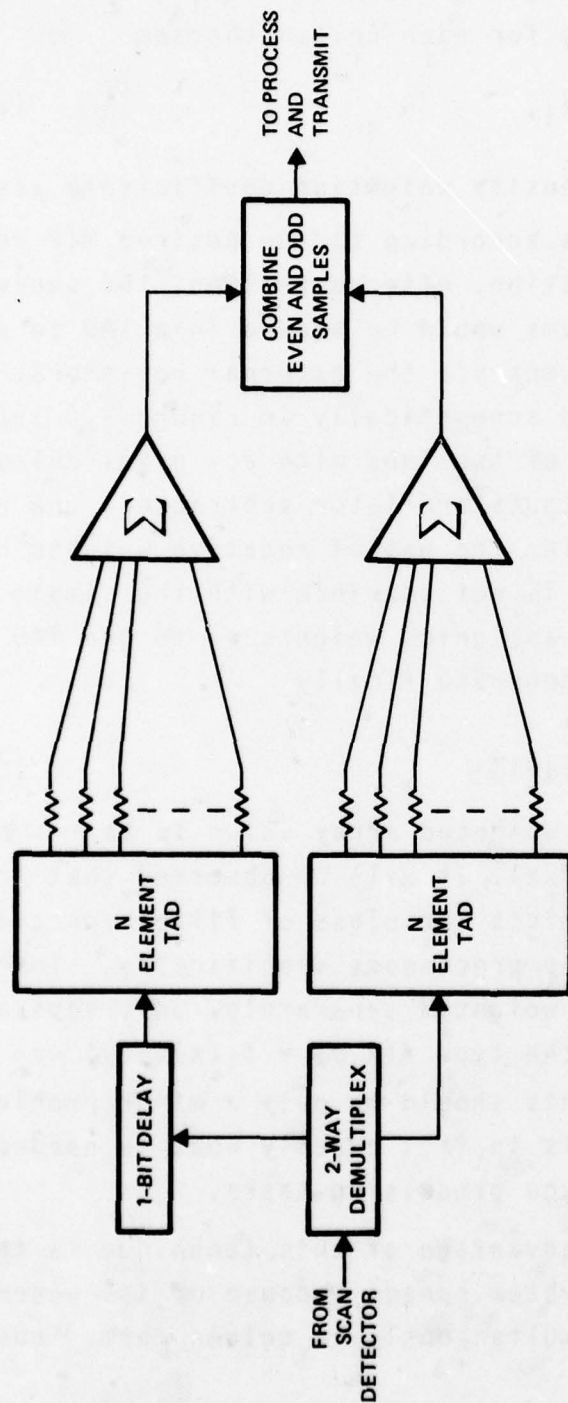
where the  $c_j$  are intensity weighting coefficients assigned to the input beams according to the desired MTF response in the vertical direction. After detection, the succession of weighted column sums would be stored in a TAD to permit weighting of the columns via the external resistors. This situation is depicted schematically in Figure 2.4.1-2. Also indicated is the use of two TADs with a 1 pixel delay between them, whose outputs are later subtracted; the purpose of this is to allow for the use of negative weights on the column values, which is not possible with the simple optical detection scheme. By assigning weights  $c_j$  to the TAD taps, the summation would generate finally

$$S_2 = \sum_j \sum_i c_i c_j a_{ij}, \quad (2-40)$$

which is the desired weighted array value to be assigned to the transmitted pixel. It will be observed that this scheme somewhat restricts the class of filter functions available to the array processor; specifically, since the rows and columns are weighted separately, only separable filter functions of the type  $f(x,y) = f_1(x)f_2(y)$  can be realized. However, this should be only a minor problem for large matrices, and is in fact exactly what is needed for certain important image processing tasks.

The chief advantage of this technique is that it can support higher system speeds because of its generation of  $N$  pixel values simultaneously in column form. Thus both





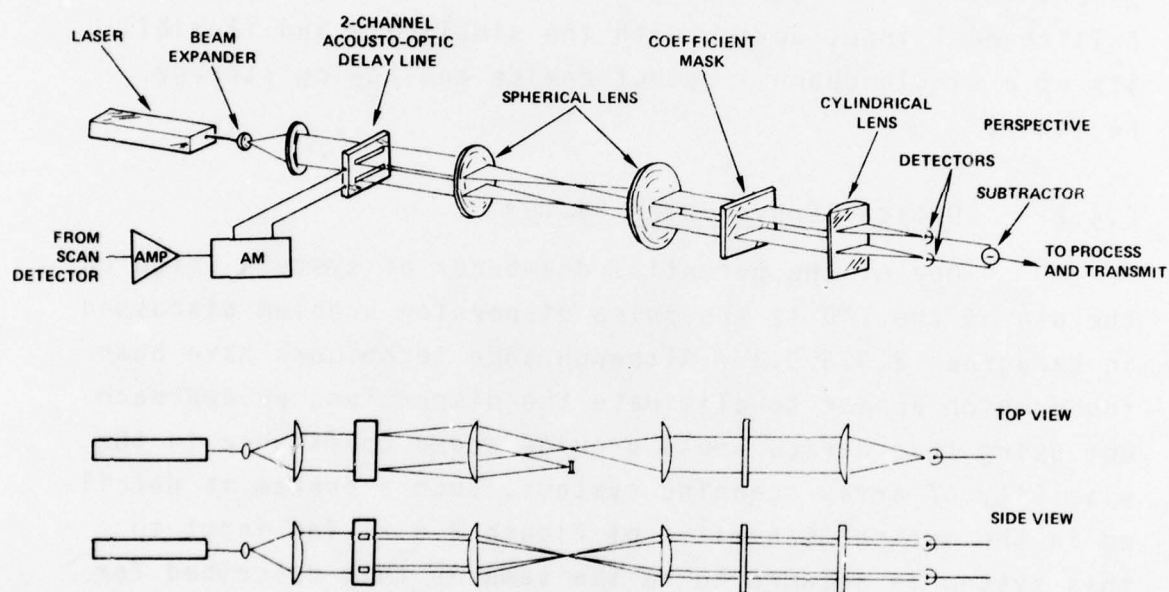
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Figure 2.4.1-2  
Column Storage And Processing Via TAD Devices

the detector and the TAD are operated  $N$  times slower for a given array size than was required in the previous system. The need for an AOB D has been eliminated, but it has been replaced by another complex device, the acousto-optic page composer(AOPC) which generates the multichannel scanning beam. However, current AOPC designs are easily capable of supporting the required rates and array sizes. In summary, the weighted linear array scanner appears to provide a good compromise design, combining the speed advantages of a multichannel input device with the simplicity and flexibility of a single channel output device and analog storage register.

#### 2.4.2 Optical Convolution System

One of the potential drawbacks of systems based on the use of the TAD is the pulse dispersion problem discussed in Paragraph 2.3.3.3.3. Although some techniques have been found which appear to eliminate the dispersion, an approach not using this device would provide added confidence in the viability of array scanning systems. Such a system is detailed in the optical schematic of Figure 2.4.2. The input to this system is assumed to be the same as that described for the dithered raster system in Paragraph 2.3.3.2; i.e., the serial pixel values from each  $N$  pixel column across the scan. In the new approach, the TAD has been replaced by an acousto-optic delay line; after modulation onto an RF carrier, the pixel values are used to drive the transducer of an acousto-optic device. Because of the finite propagation time of the acoustic waves down the crystal, many values can be present in the crystal simultaneously. When the  $N^2$  pixel values representing an array to be processed are present in



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Figure 2.4.2  
Convolutional System For Temporal  
Transversal Filtering



the crystal (and hence in the light exiting the crystal), they are imaged onto a coefficient mask where each is effectively multiplied (weighted) by the local transmittance of the mask. Then the weighted values are collected and summed in a single detector. As in the previous system, to obtain negative weights, we perform the above procedure in two channels, using one channel of the mask to provide the positive weights and the other the negative weights. The two channels are then electrically subtracted to produce the processed output.

The key tradeoff of a system of this type is the capacity of the acousto-optic delay line. For practical acousto-optic materials, the minimum acoustic velocity is about  $0.6 \text{ mm}/\mu\text{s}$ . Also, a maximum practical crystal length of about 60 mm is currently available. Thus about 100  $\mu\text{s}$  of delay can be achieved, which limits the storage in an AO delay channel to about 100 times the pixel time (e.g., at 1 Mp/s, to  $N \leq 100$ ). For the IITS requirements, this would restrict the array size to around  $11 \times 11$  when full data link rates are needed. Other systems, however, may have requirements which make this technique compatible with larger arrays. Another potential method of achieving larger arrays is to increase the number of channels in the AO delay device. For example, a very feasible device of only 8 channels would support array sizes up to around  $19 \times 19$  even at the IITS speed and resolution values. In summary, a system based on optical convolution can provide most of the advantages of the previous systems while eliminating the need for the TAD device.

### 3.0 ARRAY RECORDING

In the following paragraphs we analyze the operating parameters of several techniques suitable for array recording. The five techniques that have been identified are two approaches to multifrequency acousto-optic beam deflector (AOBD) recording, scanned AOBD recording, multitransducer acousto-optic page composer (AOPC), and LED array recording.

The two major motivations for array recording are (i) the ability to record block-encoded data without the requirement for large buffers and (ii) speed. Scanning and recording systems generally are required to be capstan-driven raster systems rather than drum devices except in those circumstances permitting the inconvenience of drum systems. A galvanometer reflector, typically used for the raster deflection, is limited to about 500 K pels/s (or 40 seconds to record a 9" x 9" image at 500 pel/in). By simultaneously recording  $N$  lines for each galvanometer sweep the recording time can be reduced by  $1/N$ .

#### 3.1 AOBD MULTIFREQUENCY RECORDING

An acousto-optic beam reflector is most commonly used as a spot scanner, a mode to be discussed in the next paragraph, by inducing into the acoustic medium a single, but changing, frequency. An alternate mode is to drive the AOBD with a multiplicity of frequencies, individually modulated so that a block of  $N$  pixels in the image plane can be recorded in parallel. In principle this can be accomplished in two ways. These are briefly described below separately to enumerate their individual characteristics. The data rate and block size analysis that follows applies to both multifrequency approaches.

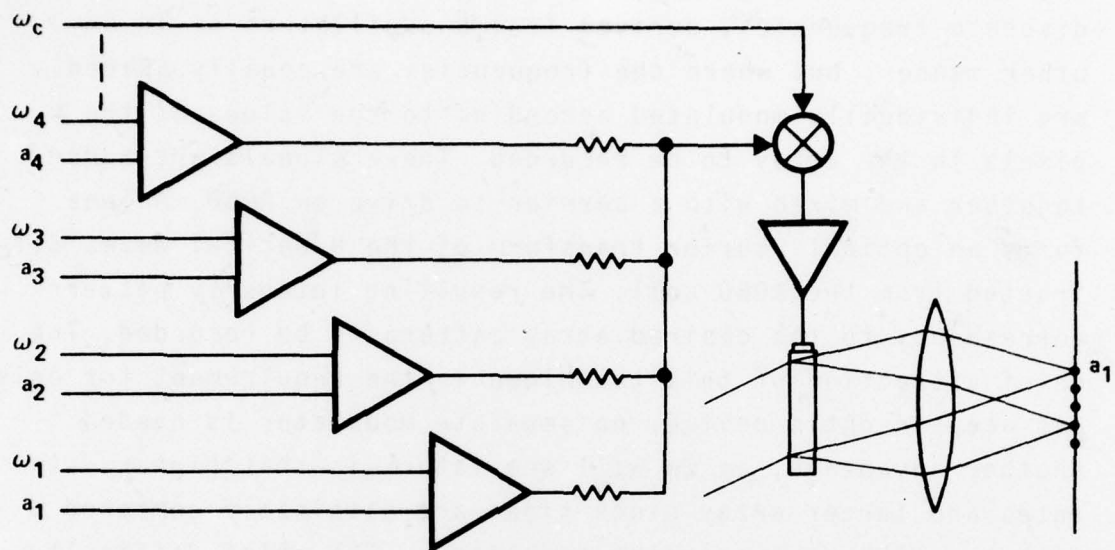
### 3.1.1 Multiple Oscillator Technique

Figure 3.1.1(a) illustrates the basic approach, in which  $N$  discrete frequencies, derived from  $N$  oscillators or in any other manner, but where the frequencies are equally spaced, are individually modulated according to the values of the  $N$  pixels in the array to be recorded. These signals are added together and mixed with a carrier to drive an AOB. A lens forms an optical Fourier transform of the  $N$  optical waves diffracted from the AOB cell. The resulting intensity pattern corresponds to the desired array pattern to be recorded. The chief attraction of this technique is the requirement for only one acousto-optic device; no separate modulator is needed. Another advantage, as we will see later, is that higher data rates and larger array block sizes are attainable compared with the AOB spot scanning techniques. The chief difficulty is the implementation of the array of frequencies, most particularly when the array is large. Another problem is that at very high data rates pixel-to-pixel distortions arise from the coherent addition of the randomly phased pixel amplitudes in the overlap areas. At lower data rates the difference in the frequencies causes the coherent effects to wash out, leaving the incoherently added intensity patterns. However, another source of distortion is the intermodulation terms associated with the diffraction of light from a complex phase-modulated grating, such as an AOB. This can force the requirement to limit the diffraction efficiency of the device to 10% or less, in order to adequately suppress these distortions. The optical efficiency of the system is thus affected and influences in turn the size of the laser required for the system.

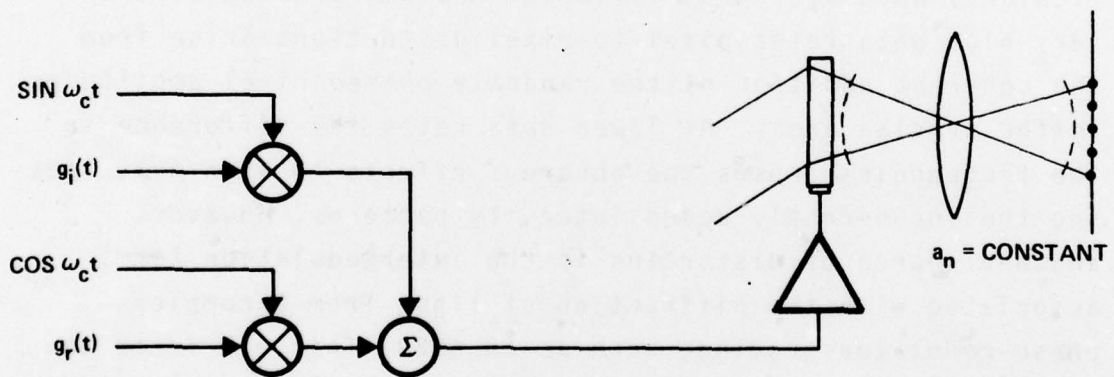
### 3.1.2 Fourier Transform AOB

An approach to multifrequency recording that requires only the carrier oscillator is sketched in Figure 3.1.1(b), where the





a. MULTIPLE FREQUENCY



b. CHIRP FOURIER TRANSFORM

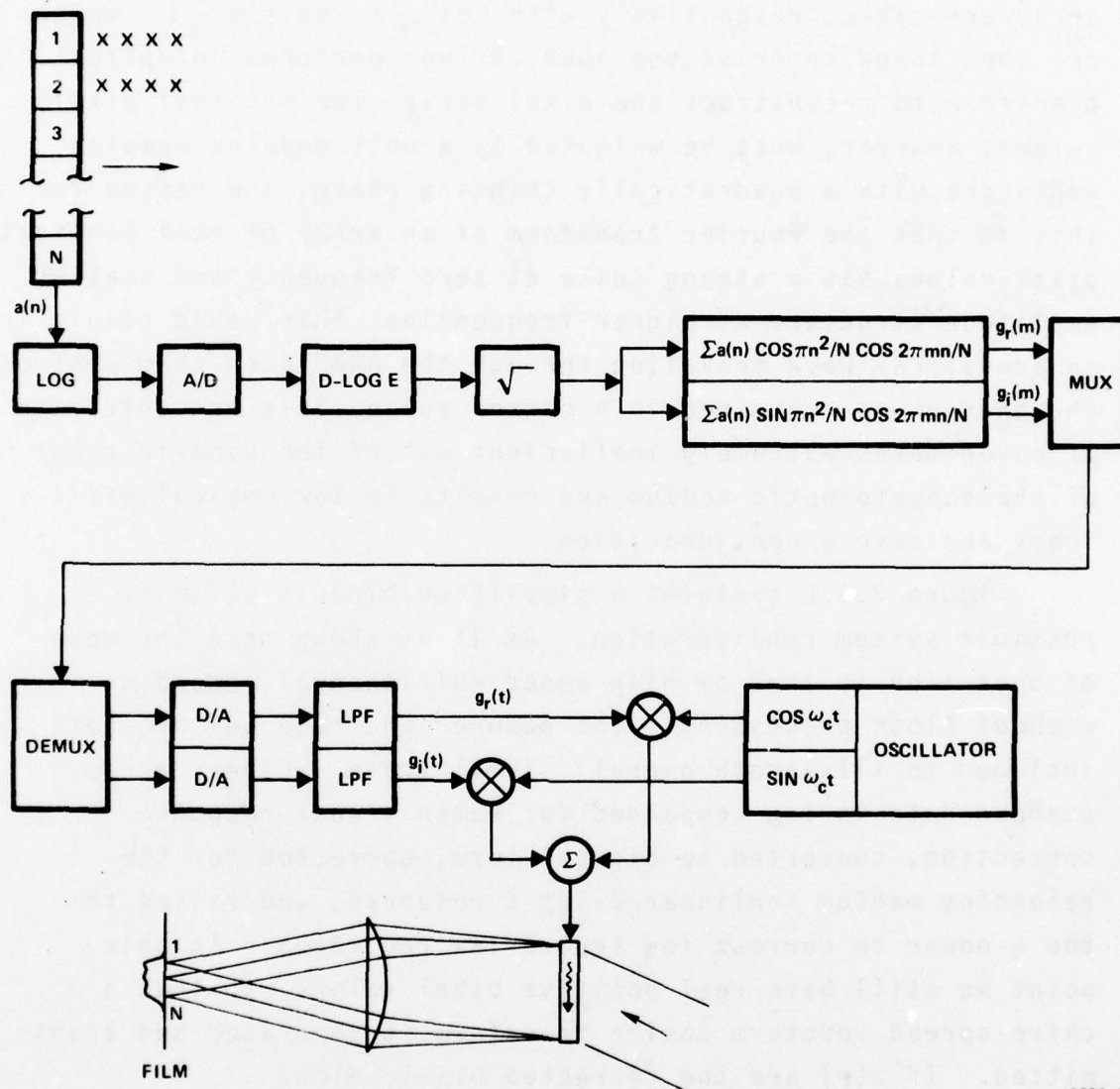
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Figure 3.1.1  
Multifrequency Techniques

real and imaginary parts of the Fourier transform of the pixel array are mixed, respectively with  $\cos \omega_c t$  and  $\sin \omega_c t$  which are then added to drive the AOBD. A lens performs an optical transform to reconstruct the pixel array. The original pixel values, however, must be weighted by a unit modulus complex amplitude with a quadratically changing phase. The reason for this is that the Fourier transform of an array of real constant pixel values has a strong spike at zero frequency and smaller amplitude structure at higher frequencies. This would result in an acoustical wave traveling through the AOBD such that most of the energy was contained in a narrow pulse. This concentration of power makes extremely inefficient use of the dynamic range of the acousto-optic medium and results in low optical efficiency and severe nonlinearities.

Figure 3.1.2 contains a simplified block diagram of a possible system configuration. As it is shown here the mode of operation is that of high speed multichannel recording without block processing. The scanner and recorder are both included to illustrate overall signal train. Linear array scanned data is log companded for human visual response correction, converted to digital form, corrected for the recording medium nonlinear D-log E response, and raised to the  $\frac{1}{2}$  power to correct for square law recording. At this point we still have real positive pixel values, so that a chirp spread spectrum cosine transform is generated and transmitted. If  $a(n)$  are the corrected pixel values

$$\begin{aligned} g_r(m) &= \sum_{n=1}^N a(n) \cos(\pi n^2/N) \cos(2\pi nm/N) \\ g_i(m) &= \sum_{n=1}^N a(n) \sin(\pi n^2/N) \cos(2\pi nm/N) \end{aligned} \quad (3-1)$$



10304-2

Figure 3.1.2  
Acousto-Optic Fourier Transform Array Recording



In the receiver the two components of the signal are converted to analog form with appropriate interpolation smoothing with low-pass filters having frequency response  $q(f)$  and impulse response  $p(t)$  to yield

$$\begin{aligned} g_r(t) &= \sum_m g_r(m) p(t-mT_o) \\ g_i(t) &= \sum_m g_i(m) p(t-mT_o) \end{aligned} \quad (3-2)$$

The signal driving the AOB is

$$s(t) = g_r(t) \cos \omega_c t + g_i(t) \sin \omega_c t \quad (3-3)$$

The refractive index of the acoustic medium is

$$\begin{aligned} n &= n_o S(t-x/v) \\ &= n_o \sum_{m,n} a(n) \cos(2\pi mn/N) \cos\left[\pi n^2/N + \omega_c(x-vt)/v\right] \\ &\quad \times p\left[(x/v) - t + mT_o\right], \end{aligned} \quad (3-4)$$

where we have used Equations 3-1, 3-2, and 3-3 and some manipulation. We will take that part of Equation 3-4 corresponding to one diffraction order to describe the diffracted light amplitude (assuming offset incident reference wave  $\exp(-ik_c x)$ ,

$$A(x) = A_0 e^{ik_c x} \sum_{m,n} a(n) \cos 2\pi mn/N$$

$$\times e^{-i\pi n^2/N - i\omega_c(x-vt)/v} p\left[(x/v)-t+mT_0\right].$$

The amplitude in the image plane is

$$B(k) = \int A(x) e^{ikx} dx$$

$$= cA_0 \sum_{m,n} a(n) \cos 2\pi mn/N$$

$$\times e^{-i\pi n^2/N} e^{ik_c vt + ik(vt-mT_0)} q(k)$$

$$= cA_0 e^{i(k_c+k)vt} \sum_n a(n) e^{i[\pi n^2/N + 2\pi(n/N)-kvT_0]}$$

$$\times \frac{\sin[(NkvT_0/2) + \pi n]}{\sin[(NkvT_0/2) + \pi n]/N}$$

$$\approx c'A e^{i\phi_0(t)} \sum_n a(n) e^{i[\pi n^2/N + 2\pi(n/N)-kvT_0]}$$

$$\times \text{sinc}\left[(NkvT_0/2) + \pi n\right]$$

$$= c'A e^{i\phi_0(t)} a(kD/2\pi) e^{i\left[\pi(kDN/2\pi)^2/N\right]}$$

where we have optically selected only positive  $k$ .

The exposure in the image plane is

$$|B(k)|^2 = K |q(k)|^2 |a(kD/2\pi)|^2,$$

which is the desired result, since the low-pass response  $q(k)$  can be made flat over the array field.

This technique, like the previous multifrequency technique, requires only a single acousto-optic device and has wider bandwidth capability than the scanned AOBBD technique. It has the additional advantage that the chirp spectrum spreading permits higher optical efficiency with lower phase distortion, especially at high data rates. For variable recording speeds, coefficients would have to be stored in a temporary memory device and clocked out at a fixed rate. Only the input rate to this memory device would vary. A question yet to be dealt with is the proper technique for pixel leveling; a precise control for this is necessary to avoid raster banding in the recorded copy, to be discussed later. Pixel leveling, required to compensate for variations in diffraction efficiency, for example, from pixel-to-pixel, must be carefully dealt with in any array scanning or recording system.

### 3.1.3 Data Rate Analysis

In analyzing the operational characteristics of a multifrequency AOBBD we will use the following quantities:

$v_a$	acoustic velocity
$D$	optical aperture of the AOBBD
$\lambda$	optical wavelength
$\Lambda$	acoustical wavelength at the center of the band



$T = D/v_a$	acoustic transit time
$f_c$	center frequency
$\Delta f$	signal bandwidth
$f_s$	sampling frequency in raster direction
$N$	number of pixels in array
$\theta$	acoustic divergence
$d$	transducer width (optical-acoustical interaction distance)
$\Gamma$	acoustic attenuation coefficient
$n$	refractive index of material

#### Acoustic Divergence Limits

The number of resolvable elements of an AOBD is given by the product of the bandwidth  $\Delta f$  of the device and the transit time  $D/v_a$ . Therefore, the resolution increases with the length and bandwidth of the device. However, as the frequency range is extended several factors combine to limit the usable optical aperture. First, higher frequencies imply shorter acoustic wavelengths, and therefore, if the thickness of the acoustic grating remains constant, a more constrained angular range of diffraction is permitted by the Bragg effect. Thus the grating thickness is made smaller (by decreasing the size of the transducer) to maintain a constant fractional bandwidth  $\Delta f/f_c$ ; in other words, the  $Q$  value is made constant, where

$$Q = \frac{2\pi d \lambda}{n \Lambda^2} \quad (3-5)$$

However, as the transducer gets smaller the divergence of the acoustic wave increases as

$$\theta = \lambda/d \quad (3-6)$$

The wave envelope spreading is usually acceptable up to the point where

$$D \leq d, \quad (3-7)$$

beyond which the wave begins to spread rapidly and the Bragg properties deviate significantly from the desired behavior. From Eqs. 3-6 and 3-7

$$D \leq \frac{d^2}{\lambda} \quad (3-8)$$

Combining Eq. 3-8 and Eq. 3-5 and noting that

$$v_a = f_c \lambda,$$

and

$$D = v_a/f_s = N v_a/\Delta f, \quad (3-9)$$

we find

$$N \leq \frac{1}{3} \frac{Q n v_a^2}{6 \pi \lambda} \frac{1}{(\Delta f)^2}, \quad (3-10)$$

where we have taken a fractional bandwidth  $\Delta f/f_c = \frac{1}{3}$ , a mildly conservative constraint to ensure no more than about 3 dB loss at the band edges.

#### Attenuation Limits

The attenuation coefficient for acousto-optic materials is proportional to some power  $m$  of frequency, generally between one and two, depending on the material. If the acoustic wave is permitted to decay 3 dB, we have

$$\Gamma f_{\max}^m D \leq 3\text{dB} \quad (3-11)$$

Since  $f_{\max} = f_c + \Delta f = (\frac{7}{2})\Delta f$ , then, using Eq. 3-7,

$$N \leq \left(\frac{2}{7}\right)^2 \frac{3}{\Gamma v_a (\Delta f)^{m-1}} \quad (3-12)$$

$\text{TeO}_2$ ,  $\text{PbMoO}_4$ , SF-8 glass, SF-59 glass and Te glass have [5, 6, 7, 8], respectively,  $m = 2, 2, 1.4, 1.25$ , and  $1.32$ . The limits for these materials have been plotted in Figure 3.1.3 along with the divergence limits of Equation 3-10. Also plotted are the  $N$  vs.  $\Delta f$  lines for values of  $f_s$  and AOB size.

### 3.2 AOBD SCANNED RECORDING

In its normal mode of operation an acousto-optic beam deflector is driven by a linear frequency ramp (or chirp), which induces a sound wave with changing frequency to propagate through the acousto-optical medium. The frequency of the acoustic wave at any instant determines the angle at which the incident laser beam is diffracted, and therefore, the position of the focused spot. A periodic chirp drive, therefore, results



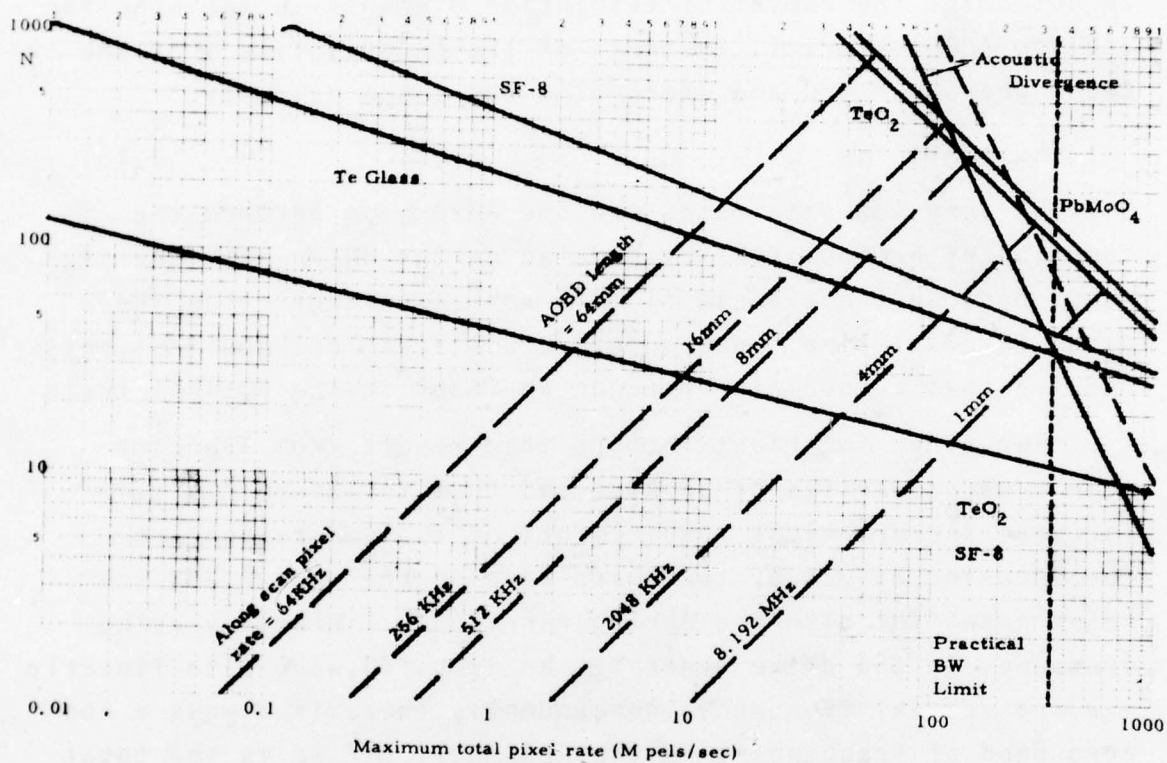


Figure 3.1.3  
Device Tradeoff Curves for an AOBD Used  
in the Frequency Mode in Array Recording Applications

in a raster scan of the focussed spot with a repetition rate equal to that of the chirp. The bandwidth  $\Delta f$  of the chirp, which determines the angular range of deflection and the distance the focussed spot travels, and the length  $D$  of the AOBD, which determines the size of the focussed spot, combine to determine the number of resolution elements in the scan for a given AOBD material. In terms of the transit time  $\tau$  of the AOBD, where  $\tau = v/D$  and where  $v$  is the sound velocity,

$$N_0 = \Delta f \tau \quad (3-13)$$

Only at very low data rates can the AOBD also perform the function of a modulator (by modulating the chirp with the signal), because the transit time is normally larger than the pixel-to-pixel time. Consequently, all practical applications require that a modulator precede the AOBD in the optical train.

Two other important factors that result from the high scan rate properties of an AOBD and that influence the performance and practical requirements of implementation are the focusing effect of the chirp wave in the medium and the requirement for blanking during retrace. A linearly varying frequency in the drive generates an acoustic wave with linearly varying spatial frequency; consequently there is always a non-zero band of frequencies in the AOBD cell. If  $\Delta f$  is the total bandwidth of the chirp, then a band  $\Delta f/N_0$  is associated with one resolution element; that is, as the range of frequencies within the cell shifts by  $\Delta f/N_0$  the focused spot shifts position by one resolution element. Furthermore, if the band within the cell changes width by  $\Delta f/N_0$ , the spot spreads over two resolution elements. This is a defocussing property that can only be corrected by optical means (e.g., altering the position of a lens). To estimate the range within which the chirp repetition rate can be varied without requiring a focus

correction, we note that the absolute value of the change in bandwidth within the cell resulting from a factor  $\gamma$  change in chirp duration  $T$  is

$$\pm \frac{T}{T} \Delta f \mp \frac{T \Delta f}{\gamma T} . \quad (3-14)$$

If  $\alpha$  is the spot growth factor acceptable (usually about 0.1), then the above expression is set equal to  $\alpha \Delta f / N$  to obtain, after some algebra,

$$\gamma = \frac{\pm 1}{\frac{\alpha T}{N^2} \pm 1}$$

Substituting for  $\tau$ ,

$$\gamma = \frac{\pm 1}{\frac{\alpha T \Delta f}{N^2} \mp 1} \quad (3-15)$$

Let us take the IITS raster rate of 256 k pels/sec, and assume we wish to maintain this horizontal scan rate (which is approaching the limit for a galvanometer in the application) and increase the recording (or scanning) data rate by an orthogonal  $N$ -pixel AOB to 256  $\times$   $N$  k pels/sec. The largest reasonable bandwidth for an AOB in this type of an application is about 250 MHz. Since  $T = 1/256$  kHz, the range of variability in the sweep time is only  $\pm 10\%$  for  $N = 32$ , and becomes even smaller for larger block sizes and/or lower chirp bandwidth. For  $N = 16$ , we find  $0.72 \leq \gamma \leq 1.6$ . For  $N = 10$  and smaller blocks the range becomes infinite on the top side (scans may be made exceedingly slow) but remain constrained for low values of  $\gamma$ . The data is summarized in



Figure 3.2-1, which shows the upper and lower limits for  $\gamma$  corresponding to several horizontal pixel frequencies, the fastest of which is that of IITS. The conclusion is that unless block sizes are very small, or slow overall scanning/recording rates are acceptable, then variable scan rates of useful range can be obtained only with optical focus compensation, and this is generally a rather difficult task, particularly if scan rates must be altered within an image.

The optical configuration for raster recording with an AOBD is illustrated in Figure 3.2-2, which includes (a) slow scan operation, (b) the defocus effect discussed above, and (c) the transition region during which the end of one chirp leaves the cell and the beginning of the next chirp enters it. During this time the modulator must be turned off, since the light distribution in the image plane consists of blurred spots at the extremes of the scan field, a result of an effectively reduced optical aperture for these frequencies. This blanking has a direct influence on the usable block size and bandwidth properties of a scanning AOBD. To analyze the data rate characteristics of this technique we observe that the useful bandwidth of the AOBD is

$$\Delta f' = \Delta f(1 - \tau/T). \quad (3-16)$$

This reduces the actual useful block size to

$$\begin{aligned} N &= \Delta f' \tau \\ &= \Delta f \tau (1 - \frac{\tau}{T}) \\ &= N_0 \beta, \end{aligned} \quad (3-17)$$

where  $N_0$  is the number of resolution elements in a scan for the limiting case of very slow scan ( $T \rightarrow \infty$ ), and

$$\beta = 1 - \tau/T \quad (3-18)$$

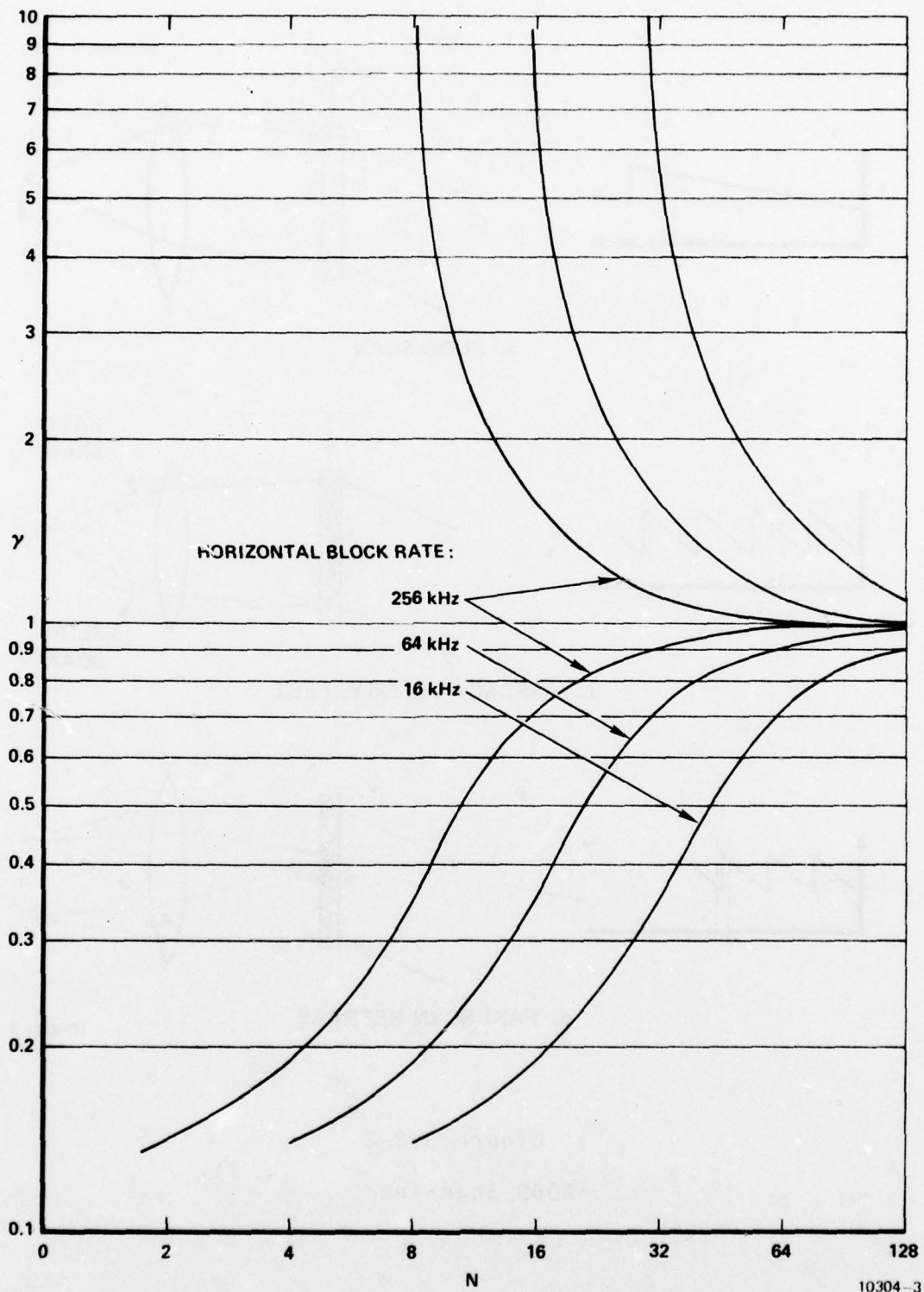
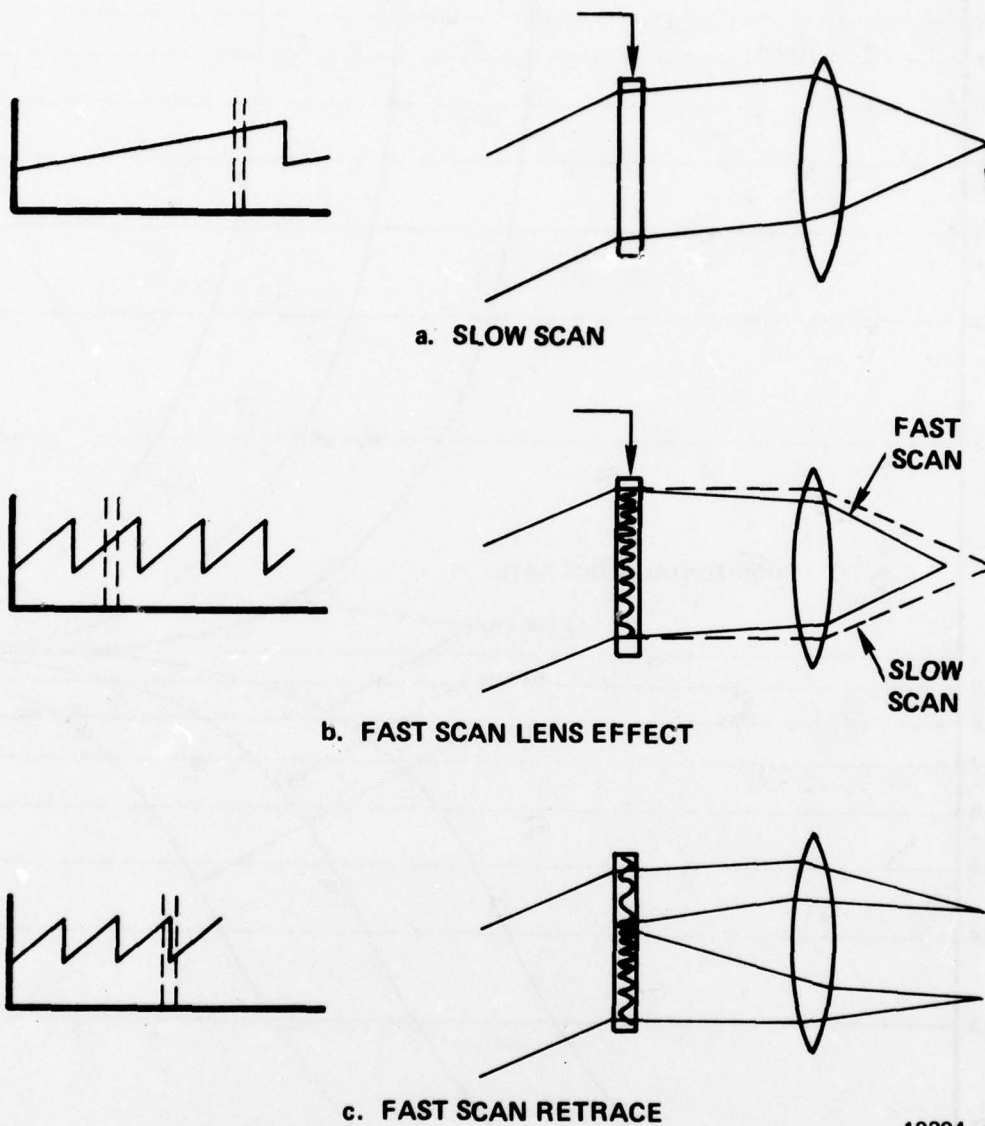


Figure 3.2-1. Chirp Rate Variation Factor Vs.  $N$



10304-4

Figure 3.2-2  
AOBD Scanning



is the duty factor of the scanner. If we assume the chirp drive retrace time is negligible, the average pixel rate is

$$\frac{N}{T} = \beta(1 - \beta)\Delta f. \quad (3-19)$$

It is apparent that the maximum obtainable data rate is  $\Delta f/4$ , which occurs when  $\beta = 1/2$ . Note, however, in this case that only  $N_0/2$  pixels are available, and that, because of the 50% blanking, a 50% exposure loss must be accepted. Figure 3.2-3 summarizes the transition constraints on an AOB. Usually it is possible and prudent to accept a lower data rate for a larger duty factor. Figure 3.2-4 shows the block size and bandwidth characteristics of an AOB scanner with the duty factor  $\beta = 3/4$  chosen to provide a reasonable compromise in the high data rate limit. The most noteworthy aspects of Figure 3.2-4 are the acoustic divergence and attenuation limits; we moved to lower frequencies and slightly down in the  $N$  coordinate compared with the curves for the AOB in a multi-frequency mode. It is evident that data rates and block rates are considerably more constrained at high data rates for recording with a scanning AOB than in a multifrequency mode.

### 3.3 MULTICHANNEL ACOUSTO-OPTIC MODULATOR (AOPC)

This technique is rather straightforward and consists simply of simultaneously modulating an array of transducers in a multichannel modulator and imaging the acousto-optical interaction plane onto the film. The beneficial aspect of this method of array recording is that it is capable of extremely high data rates. In the linear array mode the transducers can be made very small and therefore capable of very wide-band operation; as was mentioned earlier, the state of the art today is about 500 MHz of modulation bandwidth for a single channel. A multichannel device, therefore, is fundamentally

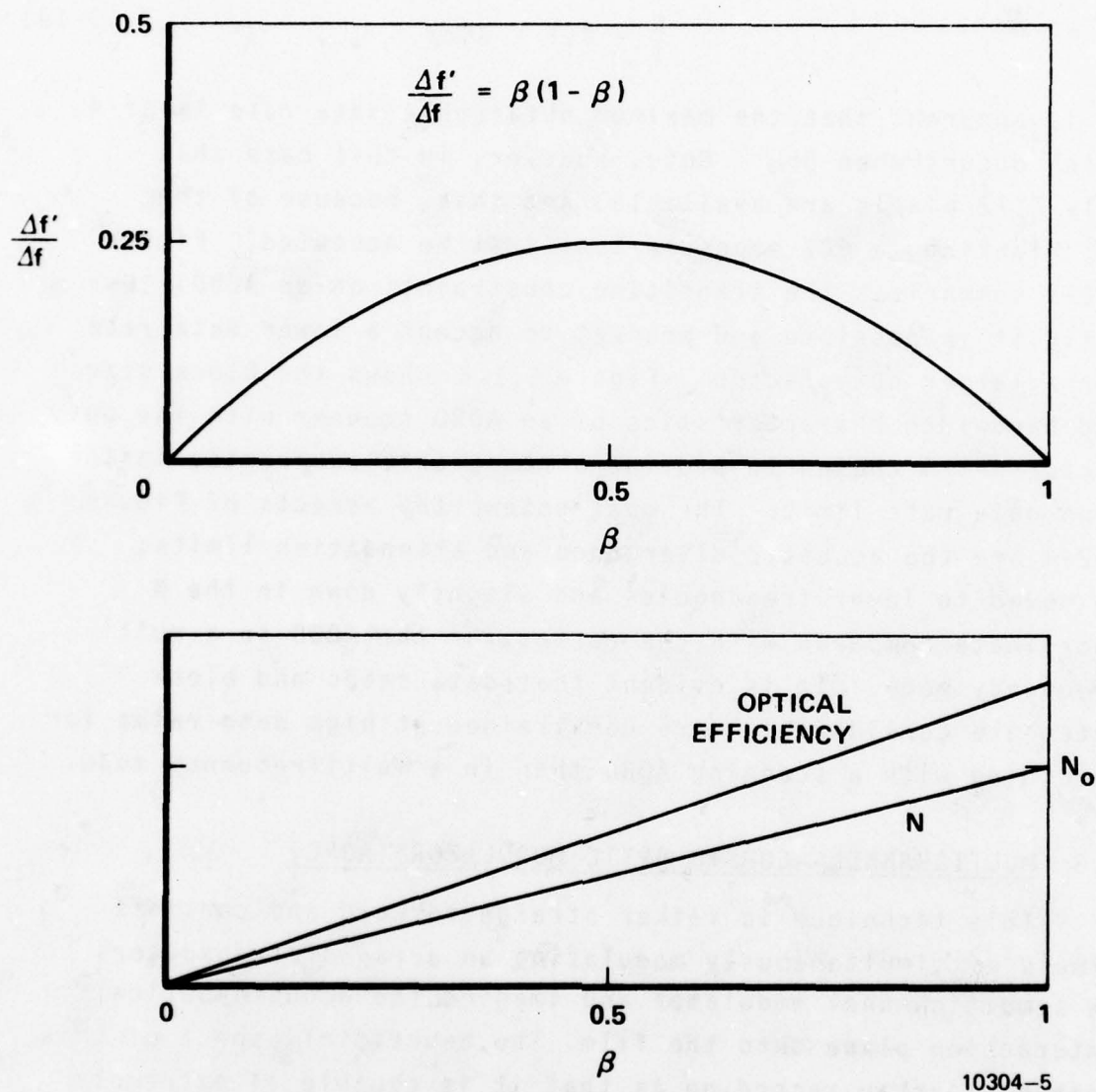


Figure 3.2-3  
Operational Parameters Vs. Duty Factor For AOBD Scanner

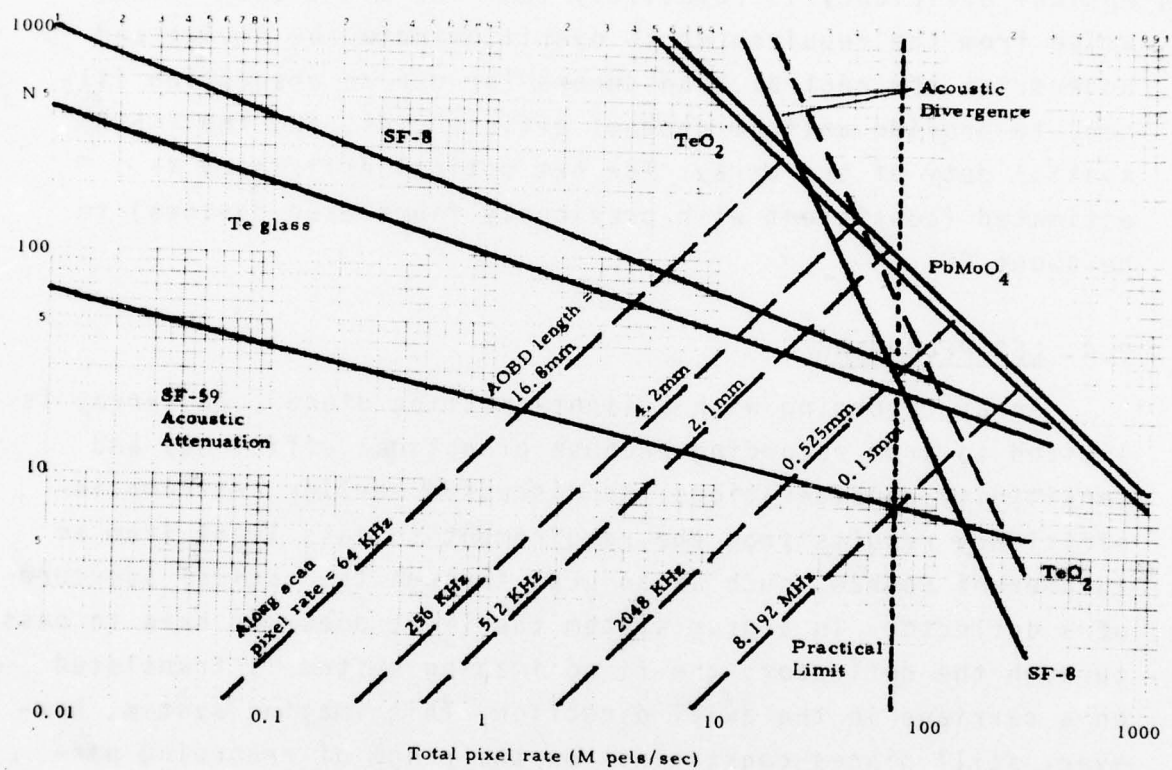


Figure 3.2-4  
Device Tradeoff Curves for an AOB D Used  
in the Scanning Mode in Array Recording Applications



capable of recording at rates far beyond any requirement envisioned. (350 MHz per channel is probably the largest practical bandwidth in this application).

The drawbacks of multichannel modulators are that they are physically complex, and therefore quite expensive, and the optical efficiency is relatively low. The efficiency losses arise from the requirement to overilluminate the device and underdrive the central transducers (or use an absorption filter) to provide uniform channel efficiencies, and the ( $\sim 50\%$ ) spatial duty of the array. The net optical efficiency is estimated (consistent with previously fabricated devices) to be about 20%.

### 3.4 LED RECORDING

Array recording with a light-emitting diode (LED) array is limited to drum recording because of optical efficiency and sensitivity considerations. As discussed earlier, extreme inefficiency results from the requirement to pass light from an incoherent source, such as an LED, through the optical aperture of a deflector. In a drum system the light does not have to pass through the deflector; the fixed imaging system is translated on a carriage in the axial direction. This imaging system, however, still places constraints on the range of recording parameters achievable.

LED array technology is still in a developing stage, oriented primarily toward visual displays. The state of the art for LED arrays applicable for the present purpose consists of single unit arrays with a maximum length of about 1 inch and a minimum center-to-center spacing of about 0.003 inch, with an 80% spatial duty factor. Larger LED element areas and spacings are readily achievable, but the 1-inch length is constrained by fabrication considerations. Larger sizes can, in principle, be

obtained by abutting smaller arrays, but this is a very expensive procedure with considerable risk in providing sufficiently accurate spacing at the abutment to avoid visual banding in the recorded image.

The light emitted from an LED can be adjusted in wavelength by suitable doping to be in the visible region (LEDs normally emit in the infra-red) to match the spectral sensitivity of recording materials. The response time of an LED is generally no more than a few nanoseconds. At 655 nm an LED emits about 12.5 mwatts/mamp/cm<sup>2</sup> up to a heat-limiting maximum of about 200 mamps of drive current. Assuming a 50% margin for diode-to-diode levelling, etc., a maximum radiant emittance of about

$$R = 1.25 \times 10^6 \text{ } \mu\text{watts/cm}^2 \quad (3-20)$$

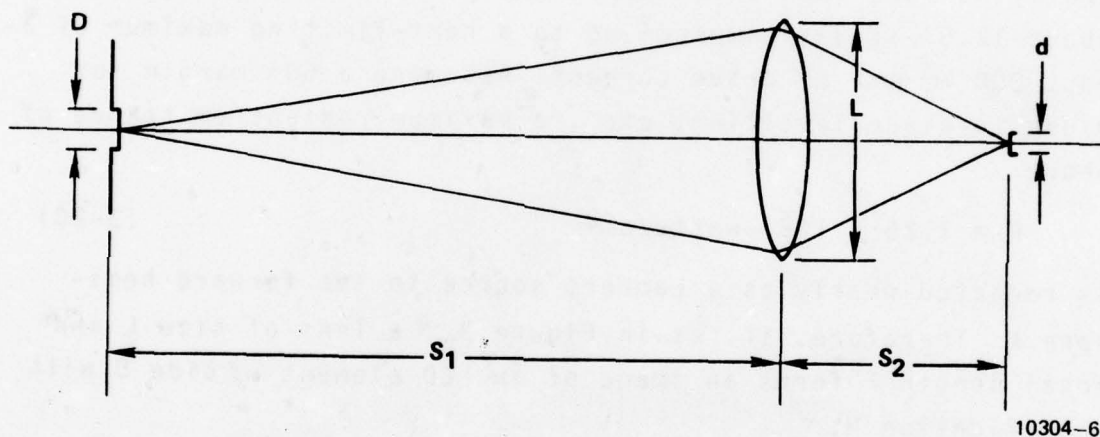
is radiated nearly as a Lambert source in the forward hemisphere. Therefore, if, as in Figure 3.4 a lens of size L and focal length F forms an image of an LED element of side D with magnification M,

$$M = d/D = S_2/S_1 \quad (3-21)$$

then the power collected by the lens is

$$\begin{aligned} P &= \frac{RD^2}{2\pi} \int_{\phi=0}^{2\pi} \int_{\theta=0}^{\theta} M \sin \theta \cos \theta d\theta d\phi \\ &= \frac{RD^2}{2} \sin^2 \theta M \\ &= \frac{RD^2}{2} \frac{1}{[1 + (2S_1/L)^2]} \end{aligned} \quad (3-22)$$

FOCAL LENGTH =  $F$



10304-6

Figure 3.4  
LED Imaging Optics



From the lens formula

$$\frac{1}{F} = \frac{1}{S_1} + \frac{1}{S_2} \quad (3-23)$$

and Eq. (3-21) we find

$$\frac{2S_1}{L} = \frac{2(M+1)}{M} (f/No.), \quad (3-24)$$

and Eq. (3-22) becomes

$$p = \frac{RD^2}{2} \frac{1}{[1 + 4(M+1)^2(f/No)^2/M^2]}. \quad (3-25)$$

The irradiance of the film is

$$\begin{aligned} I &= p/d^2 \\ &= \frac{R}{2M} \frac{1}{[1 + 4(M+1)^2(f/No)^2/M^2]}. \end{aligned} \quad (3-26)$$

Let  $r$  be the instantaneous pixel rate. The exposure of the film is

$$E = I/r. \quad (3-27)$$

The block rate is

$$\frac{r}{N} = \frac{R}{2EM^2} \frac{1}{\left[1 + 4(M+1)^2 (f/N_0)^2 / M^2\right]} \quad (3-28)$$

Let us take the example of a typical recording material used in image transmission systems, 3M type 7771 dry silver paper, requiring a maximum exposure of about  $20 \mu\text{j/cm}$ . Using this exposure and Eqs. 3-20 and 3-28, the maximum block rate for an  $f/N_0 = 2$  lens (about the minimum practical imaging lens) with magnification  $M = 0.37$  is

$$\frac{r}{N} = 1036. \quad (3-29)$$

The assumed magnification corresponds to an LED spacing of 0.007 inch reduced to the IITS line spacing of 0.0026 inch. At the IITS data rate of 256 K pels/sec, the minimum block size is

$$N = 247 \quad (3-30)$$

The size of the LED array is 1.73 inches. The number of elements in the array is too large for most applications and the physical size requires very costly fabrication. Reducing the element spacing and simultaneously increasing the magnification produces a smaller array length but a larger number of elements. On the other hand, reducing the magnification and increasing the block size means a longer array length. Reduction of the

pixel rate requirement brings the parameters within the range of practical fabrication. For example, reducing the pixel rate by a factor of ten (corresponding to a total recording time of about 10 minutes), reduces the minimum block size to 25 elements with a total length of 0.175 inch (for  $M = 0.37$ )

The performance and implementation tradeoffs involve multiple variables, primarily resolution, pixel rate, block size and array length. The general conclusion that can be made is that LED recording is limited to systems with longer recording times or lower resolution than the IITS configuration. A further disadvantage of LED array recording is that tight positioning tolerances of the surface of the recording material must be held to maintain a constant spot size, due to the strong convergence of the incoherent light. Finally, the applications are limited to those using array scanning/recording for block encoding rather than image enhancement (filtering) or high speed.

### 3.5 FIBER OPTIC LIGHT GUIDES

Fiber optics are generally most useful for transporting light through an unusual optical path, such as relaying reflected light along a line or a scanned photograph to a single small photodetector by means of a ribbon of fibers spindled at one end, or to an otherwise remotely located detector. For array recording fiber optics can be used instead of an imaging lens to bring the LED light to the photosensitive film surface. By using tapered fiber bundles (readily available from fiber-optic manufacturers), magnification and demagnification of the light spots is possible. These fiber optic links may be made somewhat more efficient than a lens system in some cases, but generally the fiber surface at the recording medium must be held



extremely close to the surface (on the order of a line spacing) at all points. This places a severe constraint on the design and film-holding tolerances that is even worse than for the lens-imaging system. Alternate techniques can generally be identified that are superior to fiber-optic coupled LED recording.

### 3.6 SUMMARY

Array recording is generally useful for high speed recording and for recording of two-dimensionally encoded data. Array scanning, on the other hand, is performed for these purposes and also to perform image enhancement operations, such as spatial filtering, which provides a single output pixel for a block of scanned pixels. For this latter operation, the recorder need only be a single line raster recorder. Although array recording is of more limited use than array scanning, it is generally less difficult to perform. There are two principal sources of difficulty in array scanning that are not present in array recording. These are the low bandwidths associated with optical detectors generally suitable for scanning applications, and the difficulty in handling efficiently the light that is diffusely scattered from a scanned object. Thus there are a number of techniques for array recording that are compatible with block sizes and data rates attainable with array scanning techniques. The considerably larger block sizes and data rates available may be used to record images stored in memory, as a means of retrieving imagery stored in a mass memory, for example, or for reducing the number of terminals required at a receiving station having large throughputs.

A summary of the principal array recording techniques that have been identified is given in Table 3.1. The AOBBD scanned recording technique seems to be the least complicated approach overall and can perform well at any pixel rate or block size

Table 3.1 Summary of Array Recording Techniques

Technique	Pixel Rate	Block Size (N)	Principal Application	Optical Efficiency	Comments
Multiple Oscillator	<350 Mpels/s	<1500	High Speed	~30%	Difficult electronic implementation. Coherent distortions at high pixel rates.
Fourier Transform	<350 Mpels/s	<1500	High Speed	~40%	High pixel rate may be expensive to implement electronically. May be directly compatible with block DFT or DCT coding. Lowest optical cost.
AOBD Scanned Recording	<75 Mpels/s	<1500	High Speed	~50%	Least electronic complexity. Lowest overall cost. Difficult to implement speed change.
AOPC Multichannel Recording	<350N Mpels/s (<500N Kpels/s for galvo)	<64	Very High Speed	~20%	Cost effective only at very high pixel rates.
LED	<250 Kpels/s	<25	Block Encoding	(~3%)	Light power & array length limited. Drum recording only.

required to be compatible with array scanning. The remaining complication, which applies to all the array recording techniques listed, is the requirement to provide optical efficiencies that are sufficiently corrected over the scan to avoid observable banding in the image.



#### 4.0 DATA COLLECTION AND PROCESSING

The data collection and processing operations that are the subject of this paragraph include resolution selection (via REARCS), image enhancement, and MTF correction. Two-dimensional filtering operations are inherent in all three of these operations and therefore have received a substantial amount of attention during the course of the study effort.

Although resolution reduction reduces the required image transmission bandwidth the selectable resolution capability is, for the purposes of this study, considered to be primarily a data collection and processing operation rather than a bandwidth compression technique and therefore is not addressed in paragraph 5.0, Bandwidth Compression Techniques.

The remainder of this paragraph deals with the theory and implementation of the two-dimensional filtering operations involved in selectable resolution, image enhancement and MTF correction and describes the results of computer simulation experiments in these three areas.

#### 4.1 THE IMPLICATIONS OF ELECTRONICALLY SELECTABLE RESOLUTION

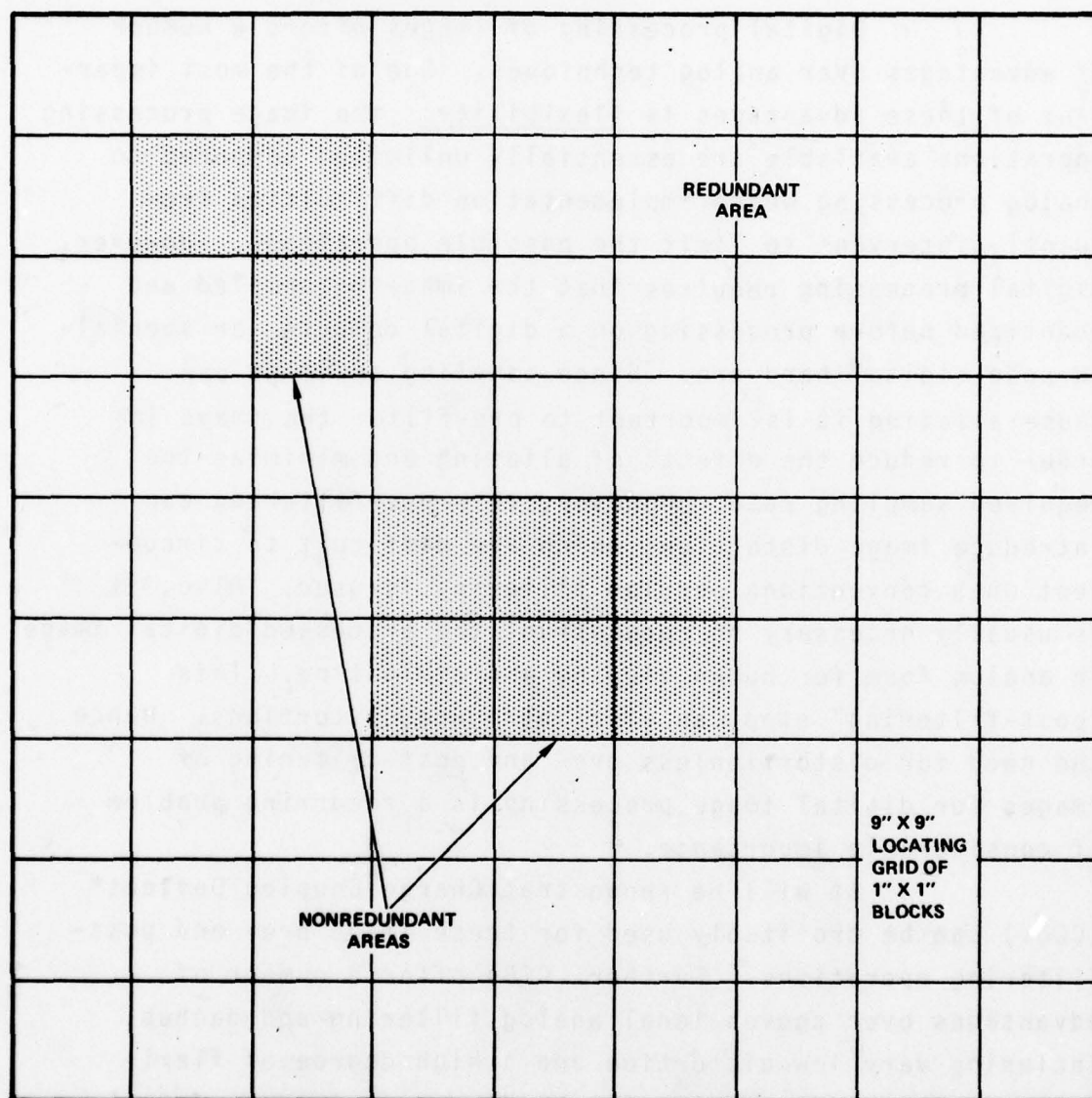
Selectable resolution is a useful feature for many image transmission systems, depending upon the end use of the transmitted images. The REARCS approach to image coding and transmission described below is particularly useful in military image systems in which some portions of an image are much more important than others for tactical reasons. If each source image is constrained to be transmitted with the same (selectable) resolution throughout the image, the desired resolution could be simply obtained by mechanically adjusting the scanning spot size and spatial

sampling rate. However, when two or more resolutions may be required within the same image, electronic resolution selection becomes very attractive for either single spot scanning systems or array scanning systems. Electronic control of image resolution has far-reaching implications; these considerations are the subject of the next section.

#### 4.1.1 REARCS and Its Impact Upon System Design

The REARCS (Redundant Area Coding System) approach to image bandwidth compression and transmission is best described by reference to Figure 4.1.1, in which so-called nonredundant or target areas, indicated by shaded blocks, are transmitted with little or no bandwidth compression at the highest spatial resolution while all remaining one-inch squares (background or redundant areas) are transmitted at a lower resolution after considerable bandwidth compression. While it is obvious that transmitting designated strips rather than designated blocks at the higher resolution would greatly simplify the system implementation, this approach would greatly diminish the attainable overall compression ratio. For such a system as REARCS in which two spatial resolutions can be present in the same image, electronic resolution reduction of high resolution imagery becomes more feasible than mechanically de-focussing a lens, especially at the higher image throughput rates. The increase in implementation complexity due to imposing a REARCS requirement will become evident in this and subsequent paragraphs.

## LOCATING NONREDUNDANT AREAS ON GRID FOR REARCS TRANSMISSION



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Figure 4.1.1  
REARCS Concept



#### 4.1.2        Distortion and Aliasing Considerations               in Sampled Imagery

##### 4.1.2.1      Introduction

Digital processing of images offers a number of advantages over analog techniques. One of the most important of these advantages is flexibility: the image processing operations available are essentially unlimited compared to analog processing where implementation difficulties frequently intervene to limit the possible operations. However, digital processing requires that the image be sampled and quantized before processing on a digital computer or special-purpose digital hardware. Since sampling an image can cause aliasing it is important to pre-filter the image in order to reduce the effects of aliasing and minimize the required sampling rate. However, this pre-filtering can introduce image distortions which are difficult to circumvent when conventional analog filtering is used. Also, it is usually necessary to reconstruct the processed digital image in analog form for human viewing and evaluating. This "post-filtering" step can also introduce distortions. Hence the need for distortionless pre- and post-filtering of images for digital image processing is a recurring problem of considerable importance.

It will be shown that Charge-Coupled Devices\* (CCDs) can be profitably used for these image pre- and post-filtering operations. Further, CCDs offer a number of advantages over conventional analog filtering approaches, including very low distortion and a high degree of flexibility. These advantages will be discussed in more detail subsequently.

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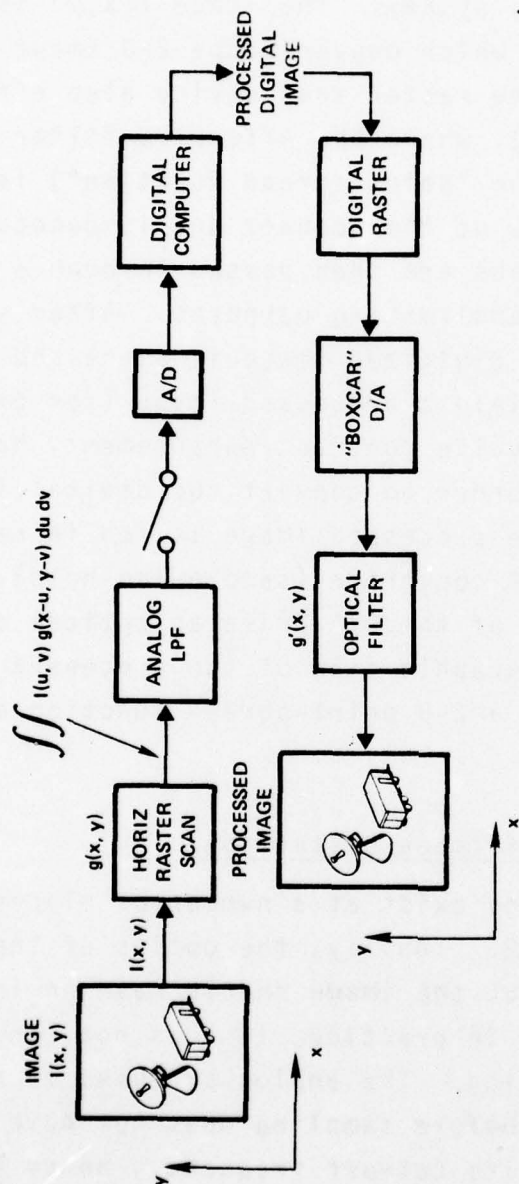
\*Here, the term "CCD" is used to denote any device which stores and transfers charge to represent sampled analog signals.

#### 4.1.2.2 Image Processing Model

Figure 4.1.2.2 shows a model for a typical digital image processing system. The image  $I(x,y)$  is scanned by a raster scan device which converts the 2-D image into a series of "lines". The raster scan device also effectively filters the image  $I(x,y)$ , where the effective filter impulse response (also called the "point-spread function") is determined by the optics of the scanner and is denoted by  $g(x,y)$ . The scanned lines are then passed through a conventional analog LPF for bandlimiting purposes. After sampling and A/D conversion, the digitized image is processed by a digital computer to yield a processed image (for example, the processing might involve contrast enhancement, bandwidth compression, etc.) In order to convert the digital image back to analog form, the processed image is fed in raster format to a "boxcar" D/A converter (sample-and-hold). The analog raster scans out of the D/A drive an optical device which generates a photographic film of the processed image. This optical device has a 2-D point-spread function denoted by  $g'(x,y)$ .

#### 4.1.2.3 Sources of Image Distortion

Distortions exist at a number of places in the system of Figure 4.1.2.2. Ideally, the optics of the raster scan device used to input the image should have an impulse point-spread function. In practice, it does not, and the result is image distortion. The analog LPF used to band-limit the raster scans before sampling does not have infinite rejection above its cut-off frequency, hence some aliasing always results from the sampling process. Conversely, it is usually desirable to minimize the sampling rate required, so the LPF is designed with a cutoff frequency



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Figure 4.1.2.2  
Model of Typical Image Processing System



which may in fact eliminate some of the image's high frequency content. In addition, the analog LPF is usually of Chebyshev or Elliptic design in order to achieve steep skirts. These types of filters have highly non-linear phase (or equivalently non-constant group delay) and poor step response, which contributes to distortion of the image and blurring of edges. After processing in the digital computer the image is reconstructed with a boxcar D/A converter, which amounts to filtering the digital samples with a sinc ( $f/f_s$ ) lowpass filter (where  $f_s$  is the sampling rate and  $\text{sinc}(x) = \frac{\sin(x)}{x}$ ). The phase of this filter is linear, but it has considerable amplitude droop and high sidelobes. The amplitude droop and high sidelobes can be compensated for by adding a compensating filter, but this will introduce phase non-linearities. Finally, the effective optical filter at the system output has a point-spread function  $g'(x,y)$  which further distorts the image by degrading the MTF.

#### 4.1.2.4 CCD Filters

CCD filters can be used to reduce these system distortions. CCD's can be used as analog shift registers to implement Finite Impulse Response (FIR) filters. Since FIR filters have linear phase and better step response than comparable lumped element analog designs, less image distortion and blurring will result. In addition, the amplitude versus frequency response of these analog shift register filters can be readily adjusted to compensate for undesirable effects of the optical filters  $g(x,y)$  and  $g'(x,y)$  in Figure 4.1.2.2. This is accomplished by adjusting the CCD tap weights to achieve the desired frequency response. Hence the CCD FIR filter offers linear phase and good step response characteristics, and can be tailored to compensate for the

point-spread distortions of the optics in the system. It can be difficult or impossible to achieve similar characteristics with conventional analog filters.. The relative merits of FIR and IIR CCD filters are discussed in Section 4.1.4.

#### 4.1.2.5 Conventional Analog Design

As an example, consider the pre-sampling filter specifications shown in Figure 4.1.2.5-1. An in-band ripple of 0.1 dB is desired, and the stop-band rejection is specified to be at least 40 dB in order to suppress the effects of aliasing. It can be shown that a 6-pole Chebyshev filter will meet these specifications at a sampling rate ( $f_s$ ) of 2.82 times the filter bandwidth (B). Figure 4.1.2.5-2(a) and 4.1.2.5-2(b) show the step response and group delay of this filter, respectively. Note the ringing in the step response, and the variation in group delay with frequency.

For image reconstruction, the system of Figure 4.1.2.2 uses a boxcar D/A. Figure 4.1.2.5-3 shows the frequency response of this boxcar reconstruction filter. The filter has linear phase (hence constant group delay) but considerable amplitude droop (roughly 4 dB at the half sampling frequency) which attenuates the high frequency content of the image. Also, the first side-lobe of the filter is down only 13 dB, which will cause the reconstructed image to be contaminated with aliased image content. While it is possible to compensate for the amplitude droop and high sidelobes by cascading a compensating filter, this filter will introduce phase non-linearities.

Note also that compensation for the point-spread functions in Figure 4.1.2.2 would complicate the conventional filter design and synthesis problem considerably.

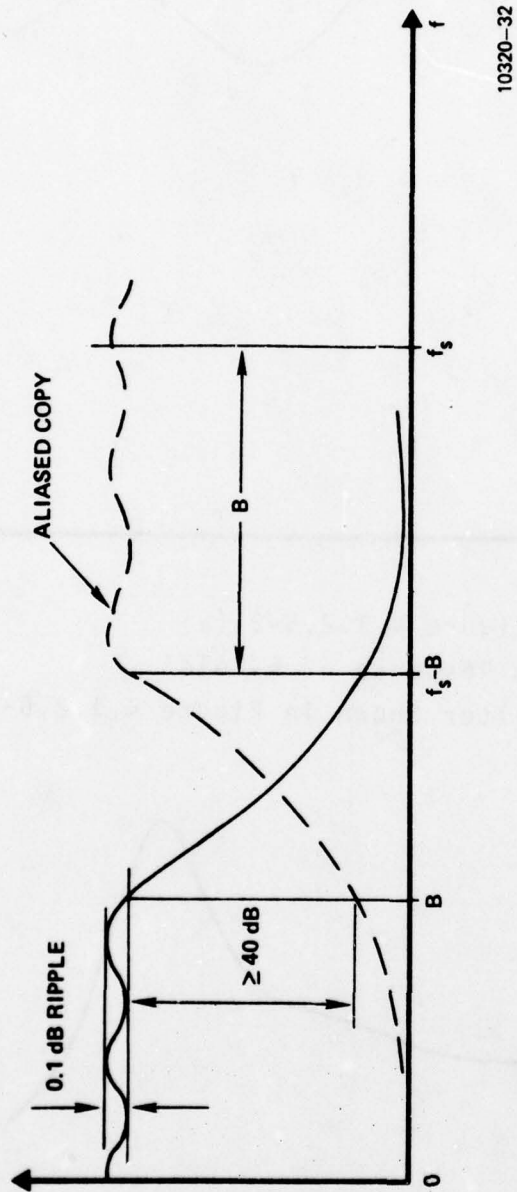
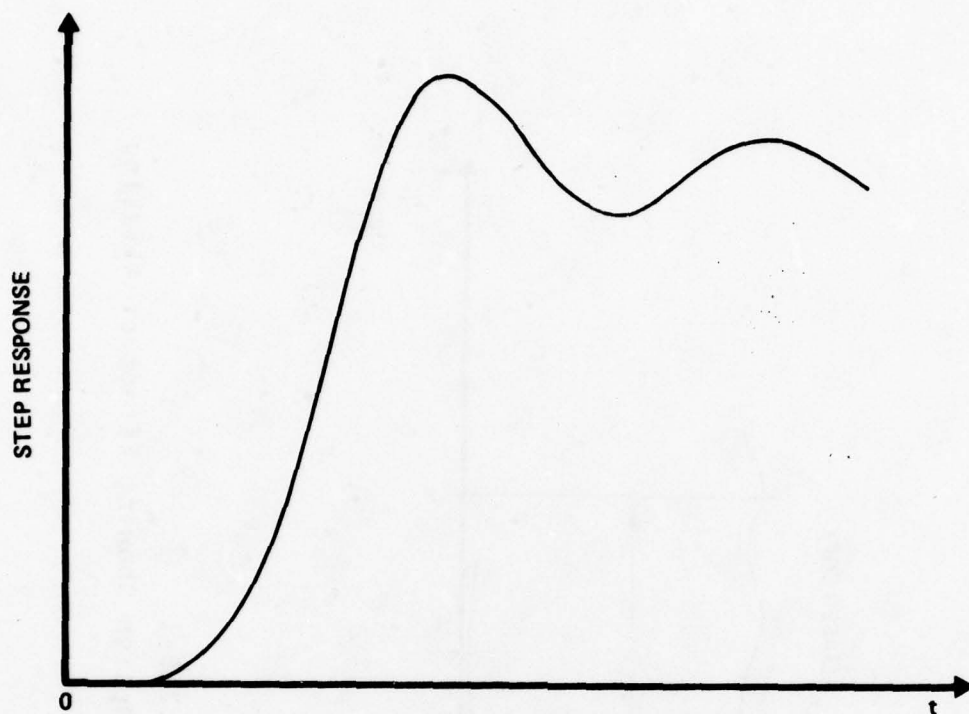


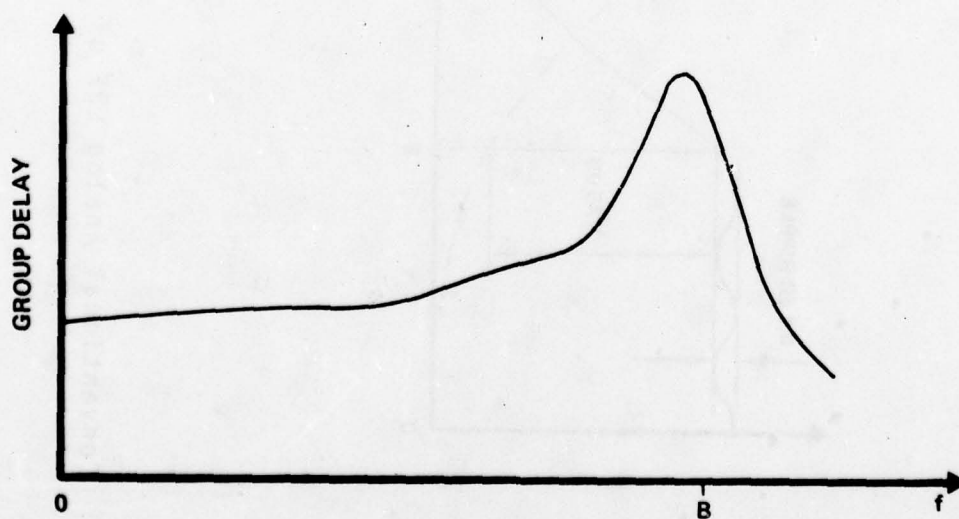
Figure 4.1.2.5-1  
Conventional Analog LPF of CHEBYSHEV Design Showing Effect of Aliasing





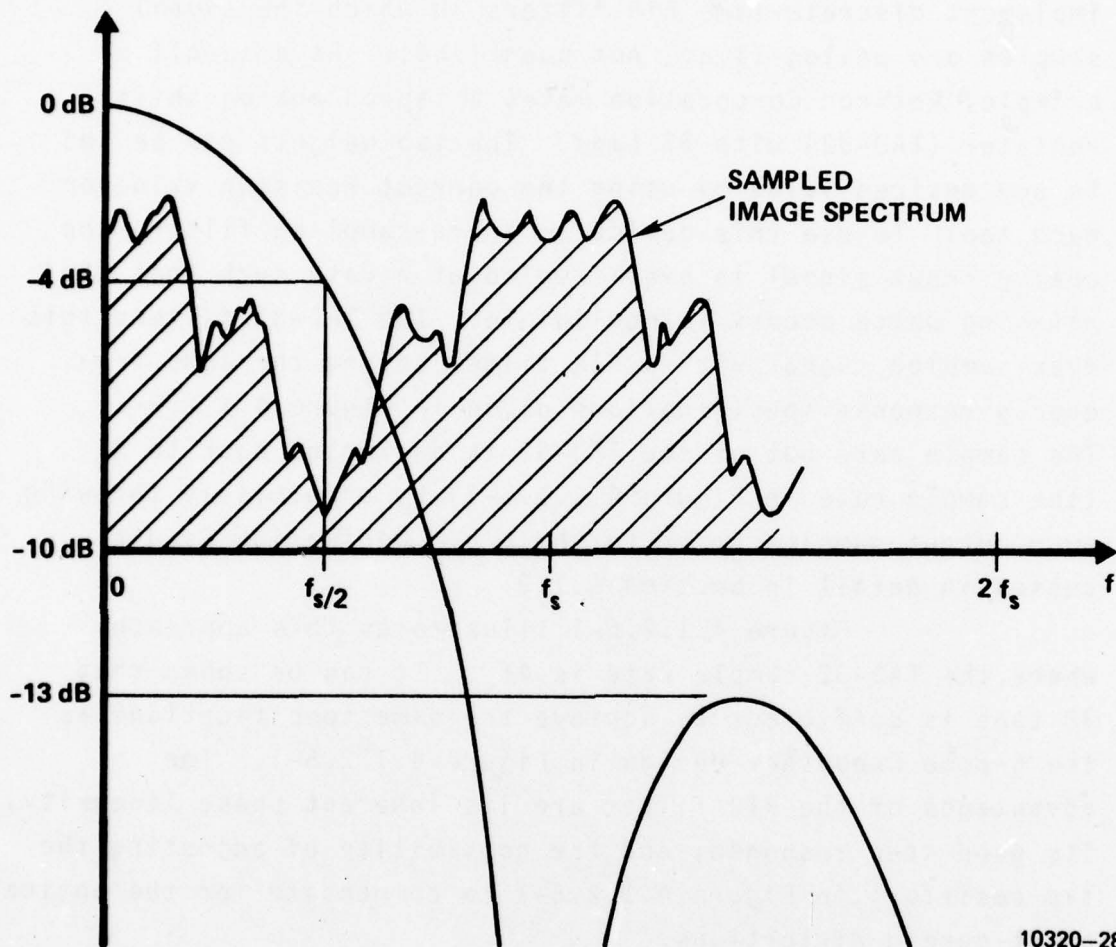
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Figure 4.1.2.5-2 (a)  
Step Response of 6-Pole  
Chebyshev Filter Shown in Figure 4.1.2.5-1



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Figure 4.1.2.5-2 (b)  
Group Delay of 6-Pole  
Chebyshev Filter Shown in Figure 4.1.2.5-1  
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Figure 4.1.2.5-3  
Frequency Response of Boxcar D/A

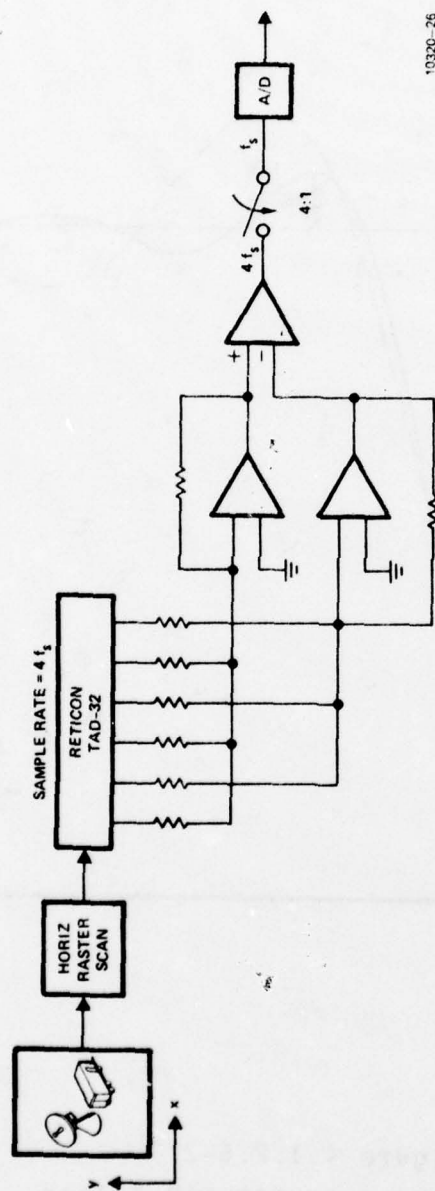
#### 4.1.2.6 CCD Design

As discussed previously, CCDs can be used to implement discrete-time FIR filters in which the signal samples are analog (i.e., not quantized). As a specific example, Reticon Corporation makes a tapped analog shift register (TAD-32) with 32 taps. The tap weights can be set to any desired value by using the correct resistor value on each tap. To use this device as a pre-sampling filter, the analog input signal is over-sampled at a rate such that the aliasing which occurs is negligible. The TAD-32 filters this over-sampled signal with a FIR filter having the same frequency response specifications given in Figure 4.1.2.5-1. The sample rate out of the TAD-32 is decimated back to  $f_s$  (the sample rate in Figure 4.1.2.5-1) by effectively throwing away output samples prior to the A/D. Decimation is discussed in detail in Section 4.1.3.

Figure 4.1.2.6-1 illustrates this approach, where the TAD-32 sample rate is  $4f_s$ . It can be shown that 32 taps is sufficient to achieve the same specifications as the 6-pole Chebyshev design in Figure 4.1.2.5-1. The advantages of the FIR filter are its inherent phase linearity, its good step response, and the possibility of adjusting the tap resistors in Figure 4.1.2.6-1 to compensate for the optical point-spread distortions.

Figure 4.1.2.6-2 shows the step response of the 32-tap FIR filter. For comparison, the step response of the 6-pole Chebyshev design is also shown. Note the smaller overshoot for the FIR filter, and its rapid approach to steady-state conditions. The group delay of the FIR filter is constant, of course, as compared to the variable group delay of the 6-pole Chebyshev filter shown in Figure 4.1.2.5-2 (b).





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Figure 4.1.2.6-1  
Use of Reticon TAD-32 For  
Pre-Sampling Filter

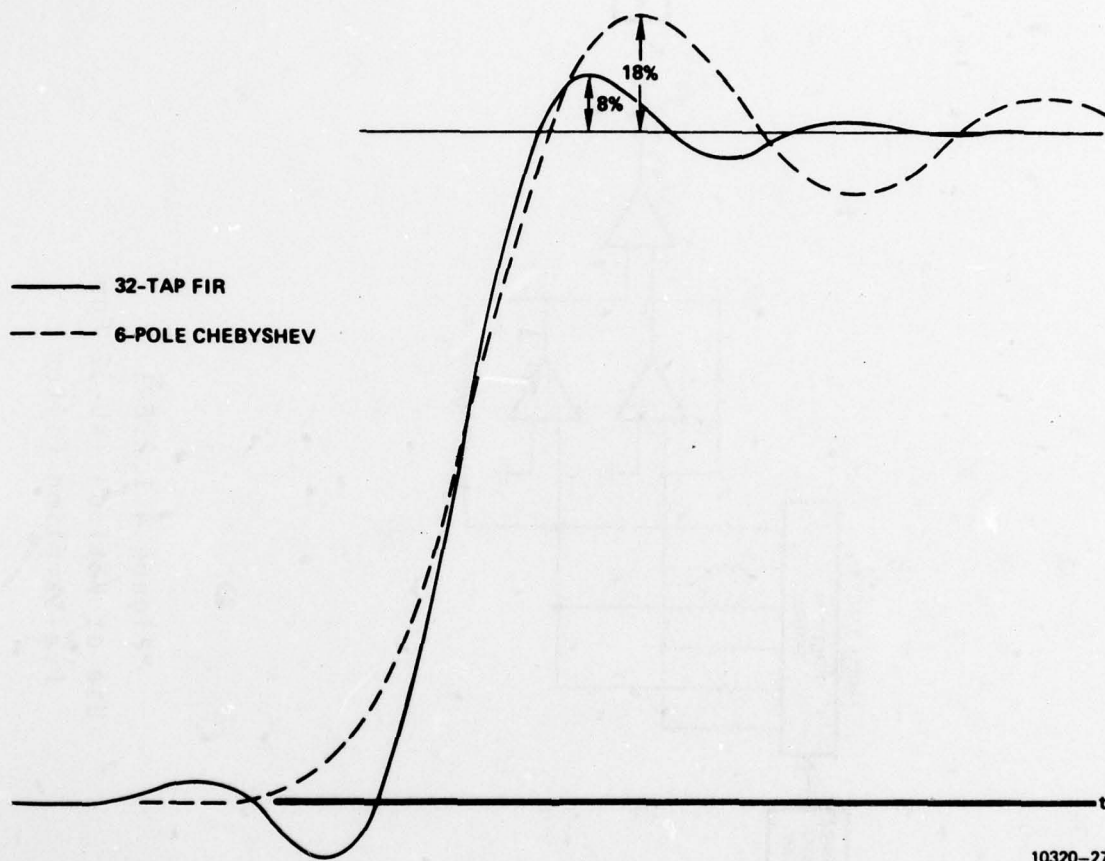


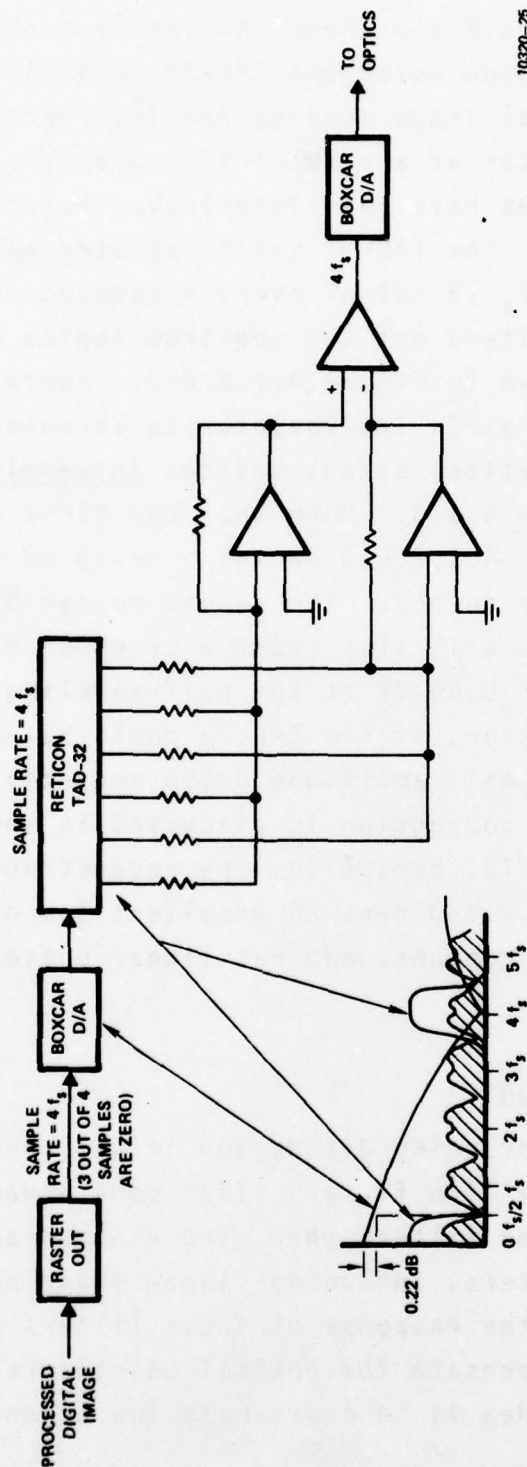
Figure 4.1.2.6-2  
Step Response of 32-TAP FIR Filter

Figure 4.1.2.6-3 shows how the processed digital image can be reconstructed using the TAD-32 to implement an FIR filter. The digital image samples are fed raster-fashion to a boxcar D/A converter at a rate of  $4f_s$  samples/second, where three zero samples have been interleaved between each pair of image samples. The TAD-32 shift register takes samples at a rate of  $4f_s$  (3 out of every 4 samples are zero). The FIR filter then filters out the spectrum copies at 0,  $4f_s$ ,  $8f_s$ , etc., as shown in Figure 4.1.2.6-3. Hence the TAD-32 is used here as a FIR interpolator to increase the sample rate into the optical output device; interpolation is discussed in Section 4.1.3. Note that the first boxcar D/A converter in Figure 4.1.2.6-3 causes a droop of 0.22 dB at the half-sampling frequency. The second boxcar D/A converter in Figure 4.1.2.6-3 will also cause a droop of 0.22 dB, yielding a net droop of 0.44 dB at the half-sampling point. Actually, the tap resistors of the TAD-32 could be adjusted to compensate for the small amplitude droop and still maintain linear phase; MTF correction is discussed in Section 4.3. With or without MTF correction, the reconstruction approach of Figure 4.1.2.6-3 does an excellent job of preserving high frequency content, and has linear phase throughout.

#### 4.1.2.7 Conclusions

In the foregoing discussion it was shown that CCD tapped analog delay line filters offer some advantages over conventional analog filters when used as pre-sampling and reconstruction filters, including linear phase and good step response. Also, the response of these filters can be easily adjusted to compensate for optical point-spread distortions. The basic idea is to oversample the scanned image





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Figure 4.1.2.6-3  
Use of TAD-32 Filter For Image Reconstruction

and then use the CCD as a FIR filter before reducing the sample rate.

While these characteristics of CCD filters seem clearly desirable on a priori grounds, the final proof of the value of these features must rest on experimental processing of images. Do the resulting images appear less distorted, blurred, noisy, etc., than conventionally filtered images? Also, the computed CCD filter responses were based upon the assumption that the CCD filters correspond to ideal FIR filters. In practice, such effects as charge transfer inefficiency, tap resistor inaccuracies, etc., will cause the CCD filter characteristics to deviate from the ideal response computed above. The deviations from ideal performance are best investigated by a combination of device modeling and analysis and experimental testing of actual devices.

The possibility of compensating for optical point-spread distortions is a very valuable feature of the CCD approach. However, this capability must also be demonstrated experimentally by measuring the point-spread function of the optics, computing the compensating filter characteristics required, implementing these characteristics in an actual CCD filter, and measuring the compensated point-spread function.

This section has served as a brief introduction to and overview of the distortion and aliasing considerations inherent in any digital image processing systems. The problem of ringing in an image due to the nonlinear phase response of presampling filters was described and a solution based upon the use of linear phase FIR sampled-data filters was proposed. The example that was presented in this section served as a convenient vehicle to introduce the important concepts of decimation and interpolation in image transmission systems. When the analog presampling filter in the example

was replaced by a FIR sampled-data filter the sampling rate was increased to prevent significant aliasing. Consequently, at the output of the FIR filter the signal frequency content was such that the samples were required less frequently, and decimation was appropriate. Similarly, the interpolation process at the other end of the channel reduced the MTF degradation due to the boxcar effect of the D/A converter. The theory of decimation and interpolation and its role in the ASITS is discussed in detail in Sections 4.1.3, 4.1.5, and 4.1.6.

It should be noted that oversampling the source image by a factor of  $D$  and then decimating the output of the presampling filter by a factor of  $D$  results in no net reduction of computational burden. The advantage lies in the fact that linear phase FIR filters can be used to minimize aliasing without introducing image distortion due to nonconstant group delay. However, if subsequent additional decimation is utilized in order to achieve electronically reduced spatial resolution, then the reduced frequency content of the decimating filter output implies that a lower sampling rate can be employed without violating the Nyquist criterion. The computational burden can then be reduced by the square of the decimation ratio without degradation to system performance. For the decimation ratios considered for the ASITS, viz.,  $D=2$  and  $D=4$ , this means a four-fold or sixteen-fold reduction in required computation at these reduced system resolutions.



#### 4.1.3 Decimation and Interpolation

##### 4.1.3.1 Introduction

The need for decimation and interpolation is a direct result of having a selectable resolution capability. While selectable resolution in an imaging system could be easily implemented via mechanical or optical means if the resolution were constant throughout the image, the additional resolution flexibility required by REARCS makes electronic resolution reduction of a high resolution image the most attractive approach for obtaining multiple resolutions within a single image.

Perhaps the most straightforward approach to reducing the resolution of an image by a factor of  $n$  is to retain every  $n^{\text{th}}$  pixel of every  $n^{\text{th}}$  scan line, ignoring all others. After these retained pixels have been transmitted to the receiver, the receiver could then repeat each received pixel value  $n$  times along each scan line and repeat each such scan line  $n$  times. As might be expected, this straightforward approach yields worse image quality than more sophisticated approaches. The remainder of this section is devoted to explaining the important issues involved in decimation and interpolation of imagery.

The straightforward but trivial decimation and interpolation process described above leads to unsatisfactory results. Figure 4.1.3.1-1 shows the original image used in this study, while Figure 4.1.3.1-2 shows the result of a 4-to-1 trivial decimation/interpolation process. As can be seen, the reconstructed image is blocky as would be expected.

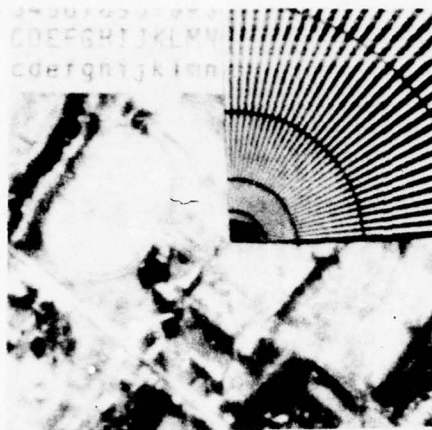


Figure 4.1.3.1-1  
Original



Figure 4.1.3.1-2  
4:1 D/I Only



Figure 4.1.3.1-3  
Decimation/Interpolating  
Using LP Filters

Figures 4.1.3.1-1, 4.1.3.1-2  
and 4.1.3.1-3

Fortunately, it is possible to generate a reconstructed image superior to that of Figure 4.1.3.1-2 by first filtering the original image before decimation and then "interpolating" between the sampled values at the reconstruction. Interpolation is defined as the generation of pixel values between the decimated values. Figures 4.1.3.1-2 uses trivial interpolation, i.e., repeating the previous point  $n$  times in each dimension. This report will investigate more sophisticated interpolation techniques using low pass filters. To illustrate that dramatic improvement is possible, Figure 4.1.3.1-3 shows the decimated and interpolated image for a 4-to-1 decimation when the correct type of decimation and interpolation is used. The details of these filters will be given below.

Basically, one must first low pass filter (LPF) the original image with a filter whose bandwidth is  $(1/n)xf_s/2$  before the decimation by  $n$  and then put  $n-1$  zeros between each output of the decimation filter ( $f_s$  is defined as the sampling frequency). This sequence of data and interspersed zeros is then put through the same LPF as was used for decimation. The theory describing this decimation/interpolation (D/I) will now be given.

#### 4.1.3.2 Decimation

The discussion to follow will be given for one-dimensional signals. All of the concepts, however, are easily extended to two-dimensional imagery. The one-dimensional analysis is given here in order to simplify the presentation. The analysis given here is taken from an excellent paper by Schafer and Rabiner [33].

Consider a signal  $\hat{x}(t)$  with Fourier transform

$$\hat{X}(\omega) = \int_{-\infty}^{\infty} \hat{x}(t)e^{-j\omega t} dt \quad (4-1)$$



Let the signal  $\hat{x}(t)$  be sampled at intervals  $T$  to generate

$$x(n) = \hat{x}(nT) . \quad (4-2)$$

The Fourier transform of the sampled data impulse stream is

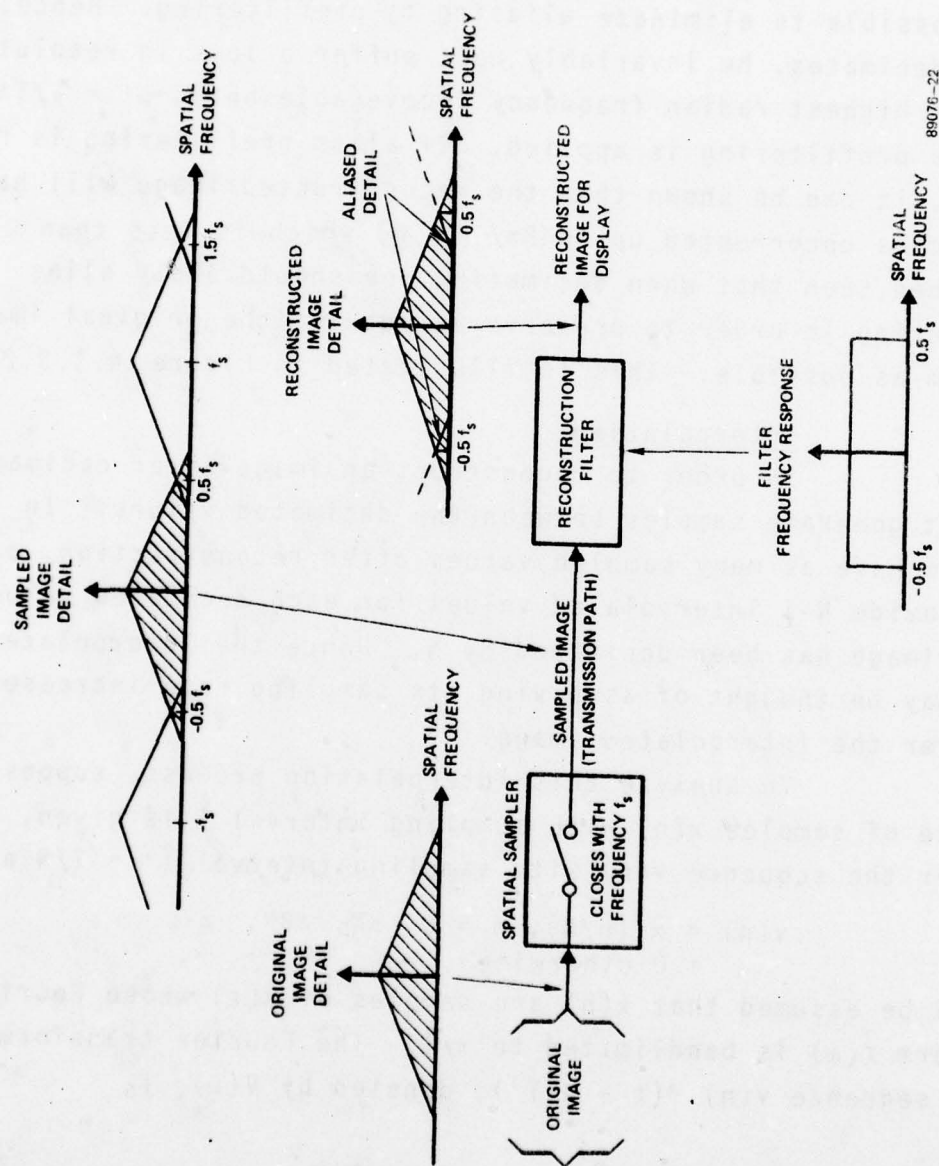
$$X(\omega) = \sum_{n=-\infty}^{\infty} x(n) e^{-j\omega nT} \quad (4-3)$$

It can be shown that

$$X(\omega) = \frac{1}{T} \sum_{k=-\infty}^{\infty} \hat{X}(\omega + k \frac{2\pi}{T}) . \quad (4-4)$$

This means that the original spectrum is repeated ad infinitum with copies of  $\hat{X}(\omega)$  repeated every  $\frac{2\pi}{T} = f_s$  where  $f_s$  is the sampling frequency. This is illustrated in Figure 4.1.3.2-1, in which it is assumed that  $\hat{X}(\omega)$  is bandlimited. Figure 4.1.3.2-1 also illustrates the important concept of aliasing. In general, aliasing occurs when the copies of  $\hat{X}(\omega)$  overlap in the repeated spectrum. If aliasing does not occur, it is possible to recover  $\hat{X}(\omega)$ , i.e.  $\hat{x}(t)$ , by using a low pass filter of radian bandwidth  $\pi/T$  [33]. Aliasing will not occur only for signals bandlimited to  $\Omega$  and provided that  $T \leq \frac{\pi}{\Omega}$ . But if aliasing occurs, it is not possible to recover  $\hat{X}(\omega)$  since  $X(\omega)$  will have aliased detail as shown in Figure 4.1.3.2-1.

Suppose that it is desirable to undersample (decimate)  $\hat{X}(\omega)$  with  $T' > \pi/\Omega$  where  $\Omega$  is the radian bandwidth of  $\hat{X}(\omega)$ . According to the above, aliasing will occur. If, however,  $\hat{X}(\omega)$  is prefiltered with an ideal LPF of radian bandwidth  $\Omega' = \pi/T'$ , then aliasing will not occur when undersampling



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Figure 4.1.3.2-1  
Sampled Imagery System With Aliasing

with  $T' > \pi/\Omega$ . It will then be possible to reconstruct the filtered version of  $\hat{x}(\omega)$  by passing the sampled values through an ideal LPF of radian bandwidth  $\Omega' = \pi/T'$ . It should be noted that it is not possible to recover  $\hat{x}(\omega)$  when undersampling but it is possible to eliminate aliasing by prefiltering. Hence, if one decimates, he invariably must suffer a loss in resolution with the highest radian frequency recoverable being  $\Omega' = \pi/T'$  if alias prefiltering is applied. If alias prefiltering is not applied, it can be shown that the reconstructed image will have frequencies uncorrupted up to  $(2\pi/T' - \Omega)$  which is less than  $\Omega'$ . It is then seen that when decimating one should apply alias prefiltering in order to preserve as much of the original image spectrum as possible. This is illustrated in Figure 4.1.3.2-2.

#### 4.1.3.3 Interpolation

In order to reconstruct an image after decimation, one must generate samples between the decimated values. In order to have as many sampled values after reconstruction, one must provide  $N-1$  interpolated values for each decimated value if the image has been decimated by  $N$ . Hence the interpolated image may be thought of as having its sampling rate increased by  $N$  over the interpolated image.

To analyze this interpolation process, suppose a sequence of samples  $x(n)$  with sampling interval  $T$  is given. Consider the sequence  $v(n)$  with sampling interval  $T' = T/N$  and with

$$\begin{aligned} v(n) &= x(n/N), n = 0, \pm N, \pm 2N, \\ &= 0 \text{ otherwise} \end{aligned}$$

It will be assumed that  $x(n)$  are samples of  $\hat{x}(t)$  whose Fourier transform  $\hat{x}(\omega)$  is bandlimited to  $\pi/T$ . The Fourier transform of the sequence  $v(n) \delta(t - nT')$ , denoted by  $V(\omega)$ , is

$$V(\omega) = \sum_{k=-\infty}^{\infty} v(k) e^{-j\omega T'k} \quad (4-5)$$



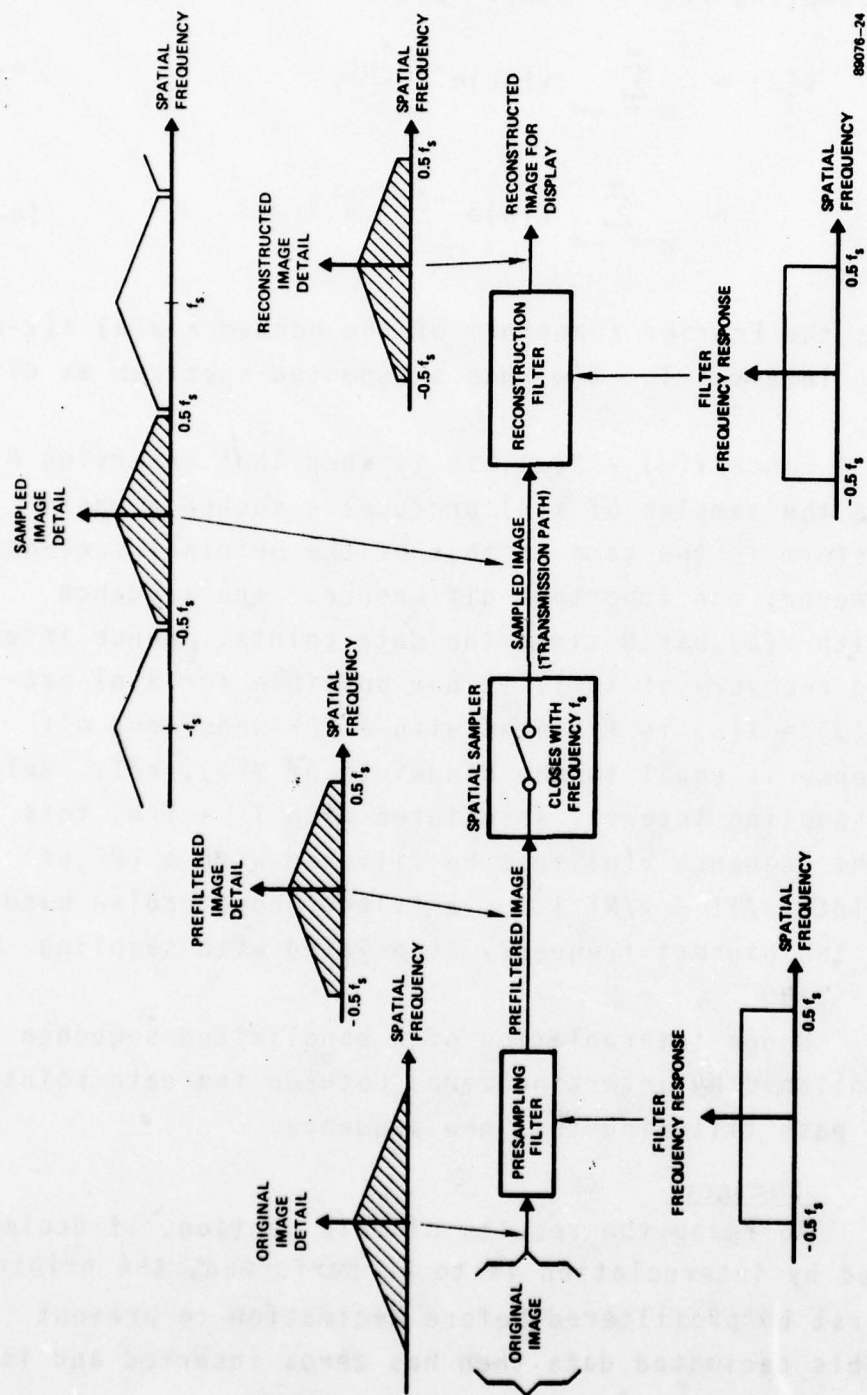


Figure 4.1.3.2-2  
Sampled Imagery System, Prefiltered to Prevent Aliasing

Note that sampling interval  $T' = T/N$  is used for  $V(\omega)$ , i.e.,  $N$  times the sampling rate of  $x(n)$ . Let  $k = nN$  and  $V(\omega)$  becomes

$$V(\omega) = \sum_{n=-\infty}^{\infty} v(nL) e^{-j\omega \frac{T}{L} nL} \quad (4-6)$$

$$= \sum_{n=-\infty}^{\infty} x(n) e^{-j\omega T n} = X(\omega) \quad (4-7)$$

where  $X(\omega)$  is the Fourier transform of the sequence  $x(n) \delta(t-nT)$  with sampling interval  $T$ .  $X(\omega)$  has a repeated spectrum as discussed above.

Since  $V(\omega) = X(\omega)$ , it is seen that inserting  $N-1$  zeros between the samples of  $x(n)$  produces a sequence whose Fourier transform is the same as that of the original sequence. There is, however, one important difference: the sequence associated with  $V(\omega)$  has  $N$  times the data points. Hence interpolation (and recovery of  $\hat{X}(\omega)$ ) is now possible for  $x(n)$  provided that  $V(\omega) = X(\omega)$  is filtered with a LPF whose cut off radian frequency is equal to the bandwidth of  $\hat{X}(\omega)$ ,  $\pi/T$ . Relative to the sampling interval associated with  $T' = T/N$ , this means that the sequence  $v(n)$  must be filtered with a LPF of radian bandwidth  $\pi/T' = \pi/NT$  i.e., a filter whose radian bandwidth is  $1/N$  the highest frequency associated with sampling interval  $T' = T/N$ .

Hence interpolation of a bandlimited sequence can be accomplished by inserting zeros between the data points, and then low pass filtering this new sequence.

#### 4.1.3.4 Summary

To recap the results of this section, if decimation followed by interpolation is to be performed, the original data must first be prefiltered before decimation to prevent aliasing. This decimated data then has zeros inserted and is

filtered by a filter whose bandwidth is equal to that of the decimation filter. In this way, the interpolated image is a filtered version of the original image. The various filtering actions and sampling spectra associated with the decimation/interpolation process are summarized in Figure 4.1.3.4.

#### 4.1.4 FIR vs IIR Sampled Data Filters and FIR Filter Parameter Tradeoffs

A two-dimensional digital filter is characterized by the two-dimensional  $z$  transform

$$H(z_1, z_2) = \frac{\sum_{i=0}^p \sum_{j=0}^q a_{ij} z_1^{-i} z_2^{-j}}{\sum_{i=0}^p \sum_{j=0}^q b_{ij} z_1^{-i} z_2^{-j}}$$

where the set of  $a_{ij}$  and  $b_{ij}$  are constants and  $b_{00}=1$ . The difference equation for this filter is of the form

$$y(n_1, n_2) = \sum_{i=0}^p \sum_{j=0}^q a_{ij} x(n_1-i, n_2-j)$$

$$- \sum_{i=0}^p \sum_{j=0}^q b_{ij} y(n_1-i, n_2-j) \quad \begin{matrix} 0 \leq n_1 \leq \infty \\ 0 \leq n_2 \leq \infty \\ 0 \neq i \neq j = 0 \end{matrix}$$



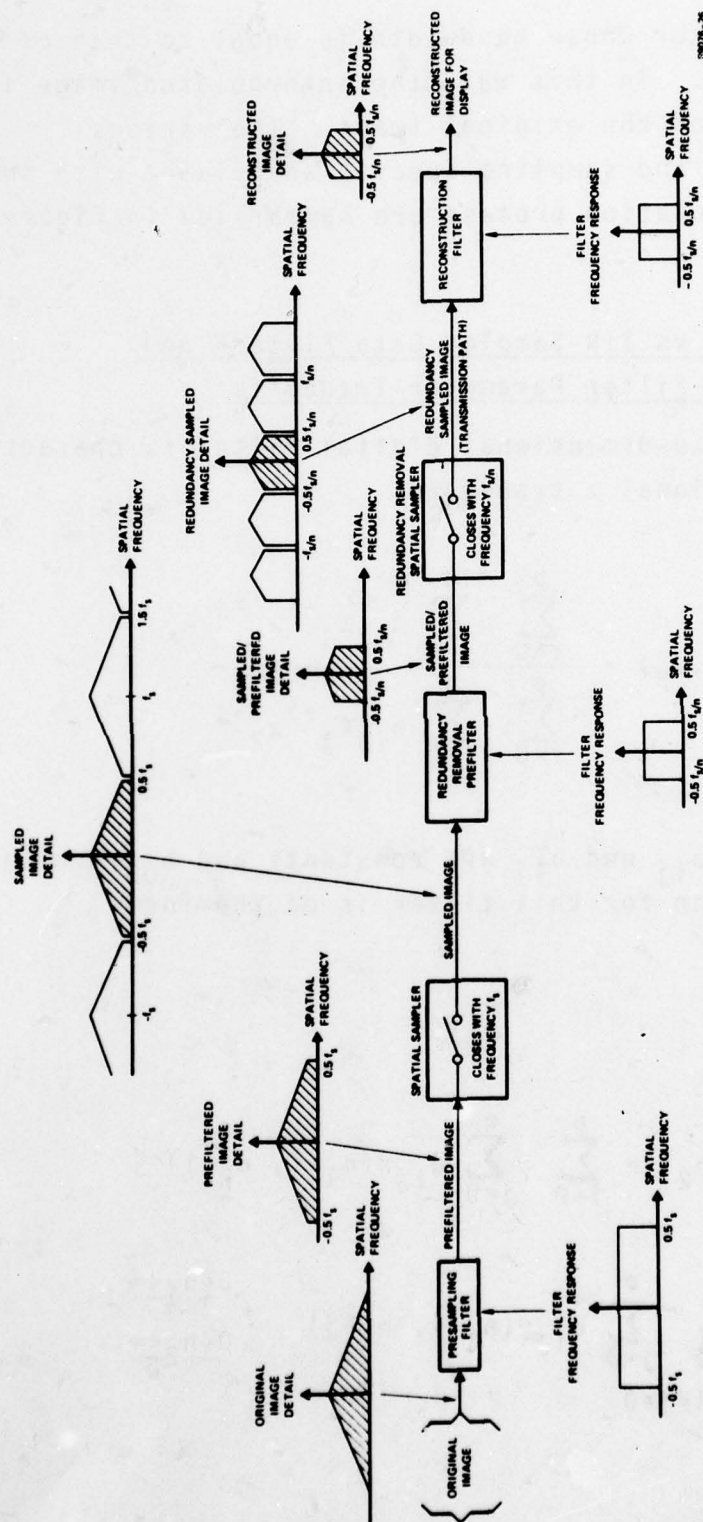


Figure 4.1.3.4

Summary of the One-Dimensional  
Decimation and Interpolation Process

where the filter recursion is in the  $(+n_1, +n_2)$  direction, i.e., the filter is casual.

In the equations above,  $a_{ij} \equiv 0$  corresponds to IIR filters and  $b_{ij} \equiv 0$  corresponds to FIR filters.

There are formidable problems associated with using two-dimensional IIR filters. These are best explained with an excerpt from Rabiner and Gold[29]:

To design infinite impulse response filter, we must be able to choose the filter coefficients  $a_{ij}$ ,  $b_{ij}$  to approximate some desired response and then we must be able to ensure that the resulting filter is stable. Both these aspects of the design problem have met with considerable difficulty to the extent that no simple design technique is readily available. It is worth discussing the sources of difficulty before proceeding to the design of two-dimensional FIR filters. As far as the filter problem is concerned, an important property of one-dimensional  $z$  transforms is the ability to factor higher-order polynomials into a product of lower-order polynomials. Thus the polynomial

$$P(z) = z^4 + \alpha_1 z^3 + \alpha_2 z^2 + \alpha_3 z + \alpha_4$$

can be factored into

$$P(z) = (z-\beta_1) (z-\beta_2) (z-\beta_3) (z-\beta_4)$$

The importance of this factorization cannot be minimized. It enables the designer to determine the stability of a high-order filter from simple criteria on the lower-order polynomials. Furthermore it provides a convenient means for realizing the filter as a cascade of lower-order filters, thereby greatly reducing certain coefficient sensitivity problems inherent in realizing high-order polynomials.

In two dimensions this factorization property is not preserved. Thus the general two-dimensional polynomial  $P(z_1, z_2)$  has no simple lower-order factors. The main implication of this inability to perform factorization is that it is almost impossible to test the stability of an IIR filter, except for some simple cases. Furthermore it implies that an unstable filter cannot be stabilized by cascading an appropriate all-pass filter with the unstable filter. The final implication is that two-dimensional IIR filters cannot be realized using a cascade configuration of lower-order filters. Thus even if one could design a stable high-order IIR filter, it would be quite difficult to realize it in an efficient way.

However, in spite of these difficulties with two-dimensional IIR filters, there is one special case involving IIR filters that could be of interest: separable two-dimensional transfer functions for which

$$H(z_1, z_2) = H_1(z_1) H_2(z_2).$$

In this case the two-dimensional transfer function can be realized using two one-dimensional filters, where the filters could be both FIR, both IIR or one of each.

The discussions in the remainder of this section will deal with one-dimensional filters which, as just pointed out, can be used to realize two-dimensional separable transfer functions. In addition, a procedure to be described later exists for generating two-dimensional nonseparable transfer functions from one-dimensional FIR filter designs.



Given the need for decimation and interpolation filters, one must evaluate the relative merits of FIR and IIR sampled data filters. Generally speaking, FIR filters have exactly linear phase and thus there is no group delay signal distortion. On the other hand, IIR filters have nonlinear phase which gives rise to signal distortion, but IIR filters are generally easier to implement than FIR filters with comparable frequency transfer function magnitude response. For the purposes of this discussion ease of implementation of discrete-time filters will be measured in terms of the number of multiplies per sample. Since the importance of linear phase in image filtering is well established [12] it is of some interest to determine the magnitude of the implementation complexity penalty that results when choosing FIR filters over IIR filters. Some valuable work in this area has been reported in [13] in which quasi-equiripple FIR filters were compared to elliptic IIR filters realized as a cascade of second-order filter sections. For equivalent FIR and IIR filter designs, i.e., both having the same values of  $\delta_1, \delta_2, F_p$ , and  $F_s$  as defined in Figure 4.1.4-1, equivalent implementation complexity is obtained when the number of IIR multiplies per sample is equal to the number of FIR multiplies per sample:

$$\left[ \frac{3n+3}{2} \right] = \left[ \frac{N+1}{2} \right] \quad (4-12)$$

or

$$\frac{N}{n} \approx 3 + \frac{1}{n} \quad (4-13)$$

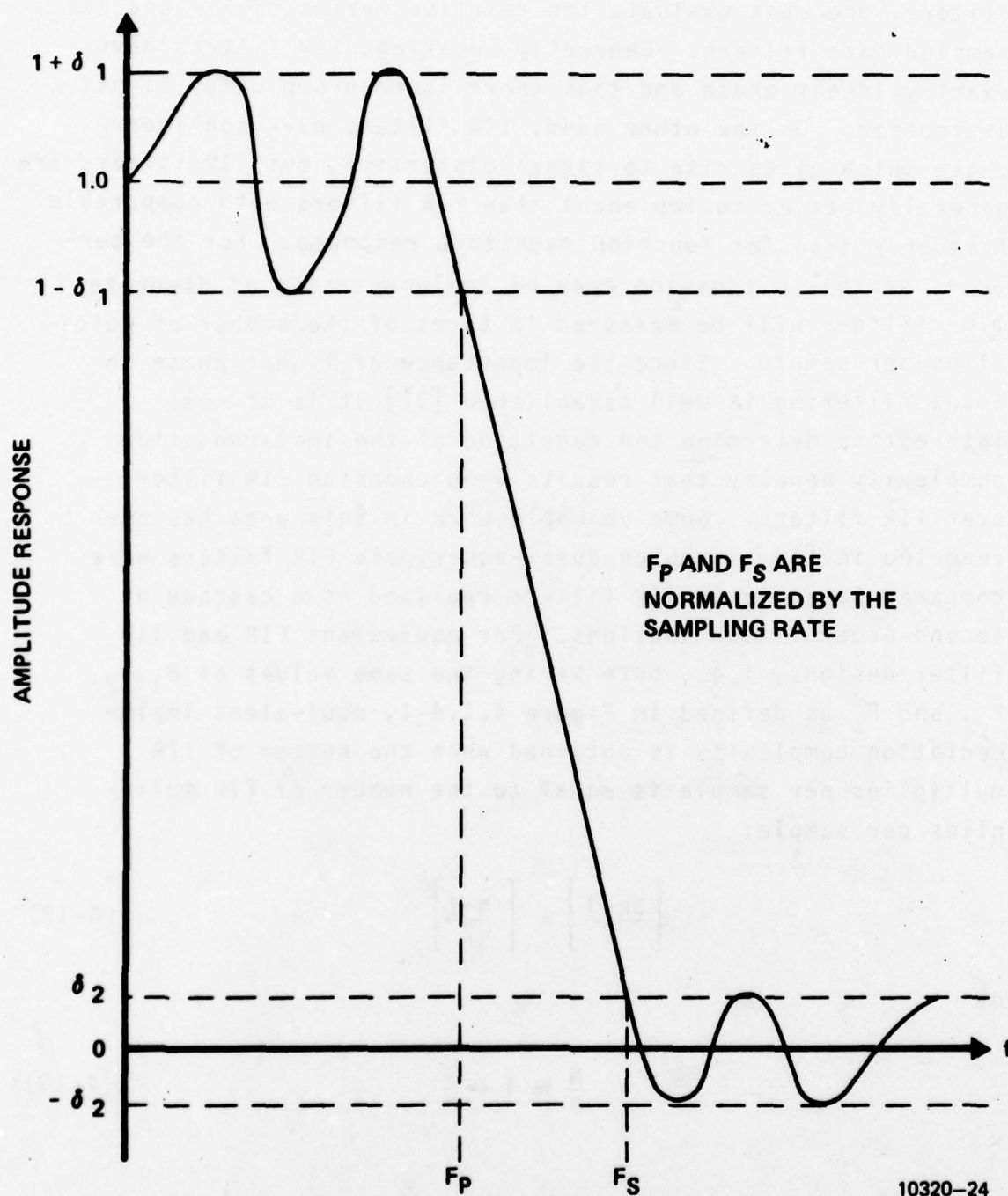


Figure 4.1.4-1  
Filter Terminology

where

$N$  = number of FIR stages

$n$  = order of IIR filter

and

$[\cdot]$  = integer part of.

Rabiner *et. al* measured  $N/n$  as a function of  $n$  for filter designs with a wide variety of values for  $F_p$ ,  $F_s$ ,  $\delta_1$ , and  $\delta_2$ . In general, it was found that large values of  $\delta_1$ , small values of  $\delta_2$ , and large transition bandwidths ( $F_s - F_p$ ) led to FIR designs that compared favorably with IIR designs in terms of complexity. More specifically, they observed that

- (1)\* For  $F_p \geq 0.3$ ,  $N/n > 3 + 1/n$  for all values of  $\delta_1$ ,  $\delta_2$ , and  $n$
- (2) For  $n \geq 7$ ,  $N/n > 3 + 1/n$  for all values of  $\delta_1$ ,  $\delta_2$ , and  $F_p$
- (3) The smaller the value of  $F_p$ , the larger the range of  $\delta_1$ ,  $\delta_2$ , and  $n$  for which  $N/n < 3 + 1/n$

For the decimation ratios of interest in ASITS, e.g., 2 and 4,  $F_p = 0.25$  and  $0.125$ , respectively, so condition (1) above implies that there exist some choices of  $\delta_1$ ,  $\delta_2$ , and  $N \approx 3n + 1$  for which FIR designs are less complex than IIR designs. Condition (2) implies that for  $N > 22$ , all FIR designs will be more complex than IIR designs, regardless

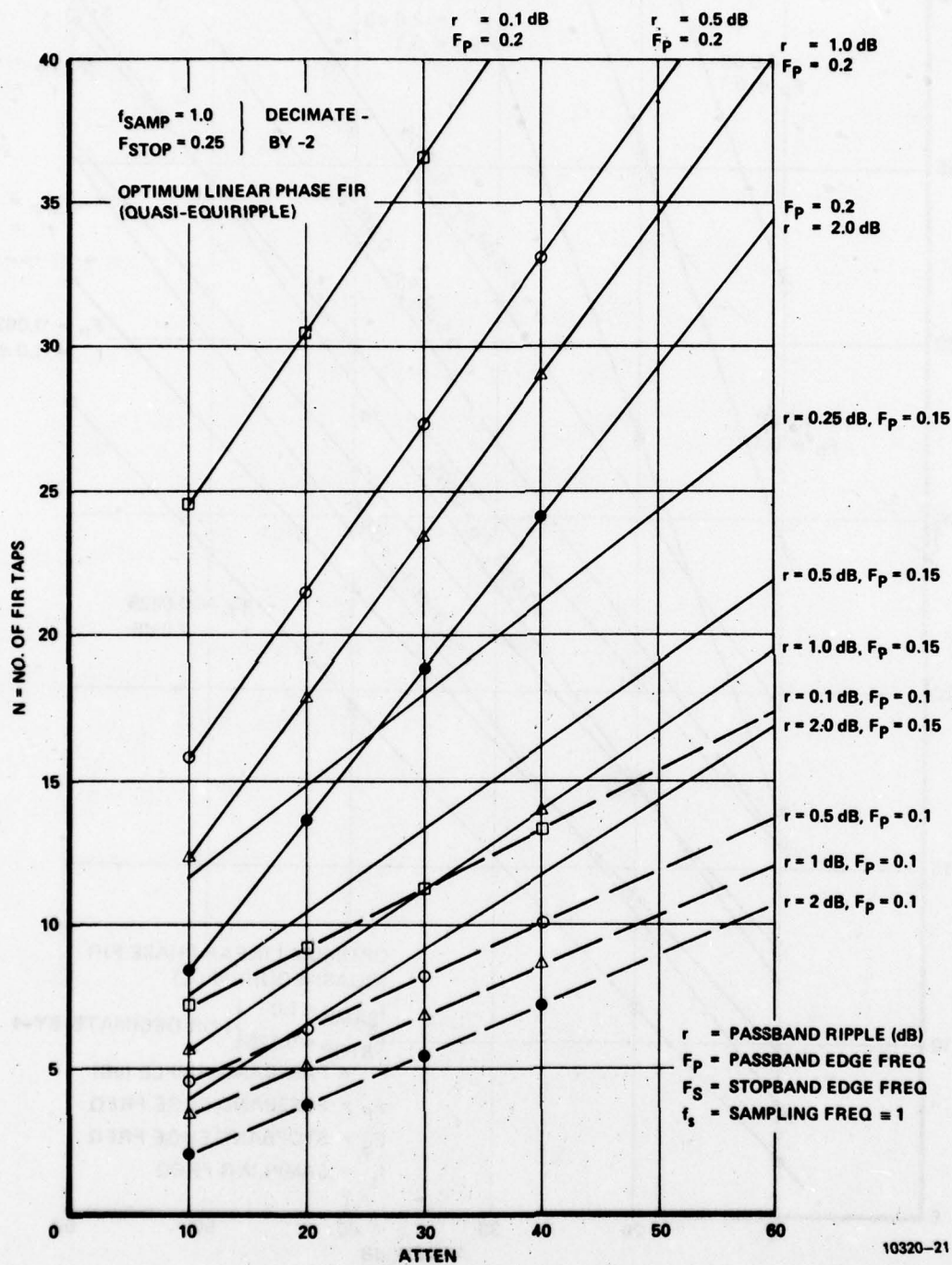


of  $\delta_1$ ,  $\delta_2$ , and  $F_p$ . Computer simulation experiment results described in Section 4.1.7 and for which  $N=9$ , 15, and 31 indicate that  $N > 22$  is highly desirable in the FIR filters used for decimation and interpolation. Consequently, it appears that the required FIR filters will require more multiplies per sample than equivalent IIR filters. However, the fact that the break-even point occurs for a value of  $N$  as high as  $N=22$  indicates that the additional FIR filter complexity compared to IIR filter complexity may not be of consequence for the values of  $N$  likely to be used in ASITS. Therefore, the advantage of having linear phase filters is obtained for a modest increase in computational complexity.

Given that FIR filters are appropriate for the ASITS application, the trade-offs among the various FIR filter parameters are of interest. Using the empirical relationship [13]:

$$N = \frac{-10 \log_{10}(\delta_1 \delta_2) - 15}{14(F_s - F_p)} + 1$$

these filter parameter trade-offs can be displayed graphically as shown in Figures 4.1.4-2 and 4.1.4-3. In both of these figures,  $\delta_1 = \delta_2$  was assumed. An alternative method of displaying these tradeoffs that yields more insight is given in Figure 4.1.4-4.



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Figure 4.1.4-2  
 Filter Tradeoffs for D/I-by-2

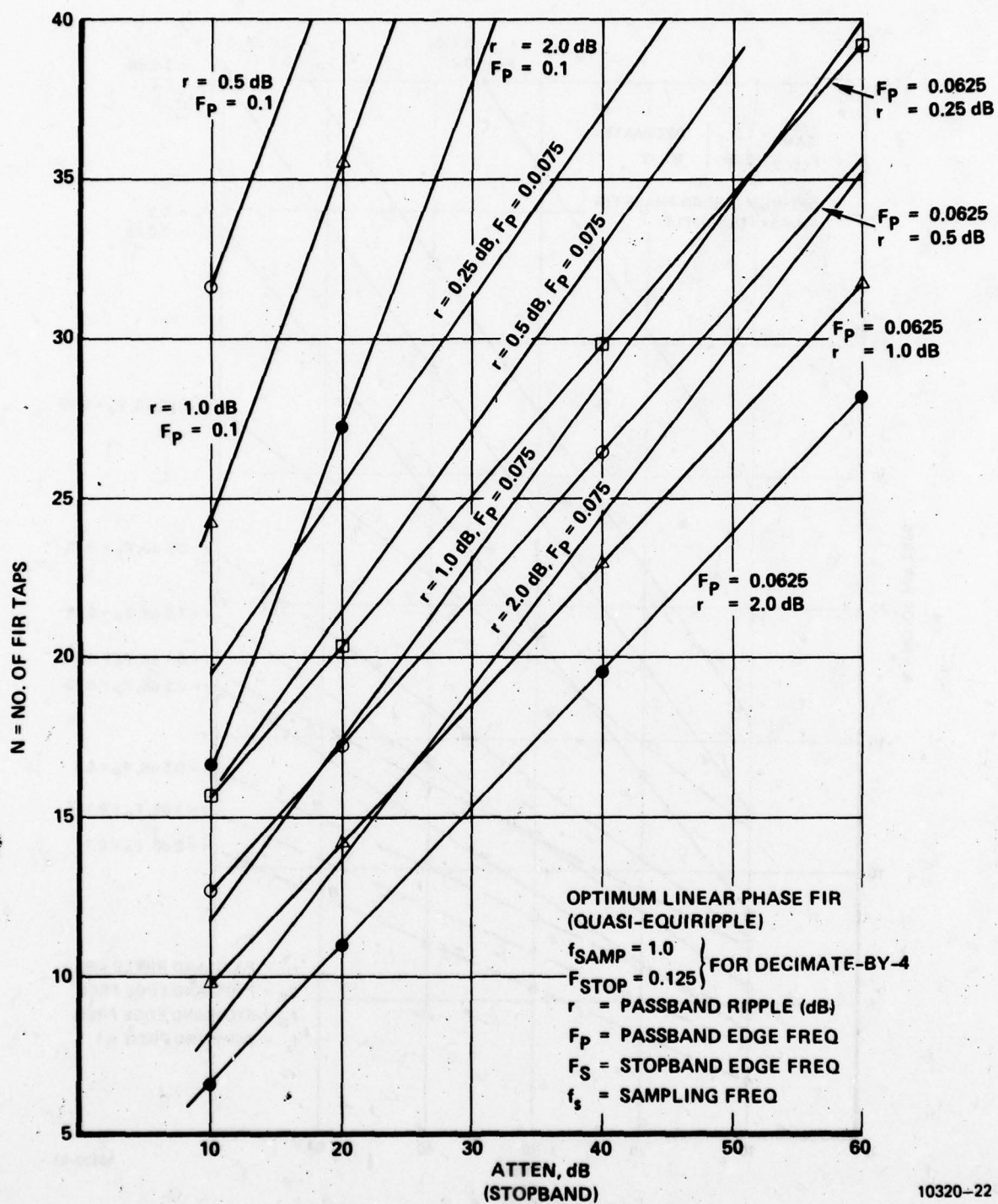


Figure 4.1.4-3  
Filter Tradeoffs for D/I-by-4



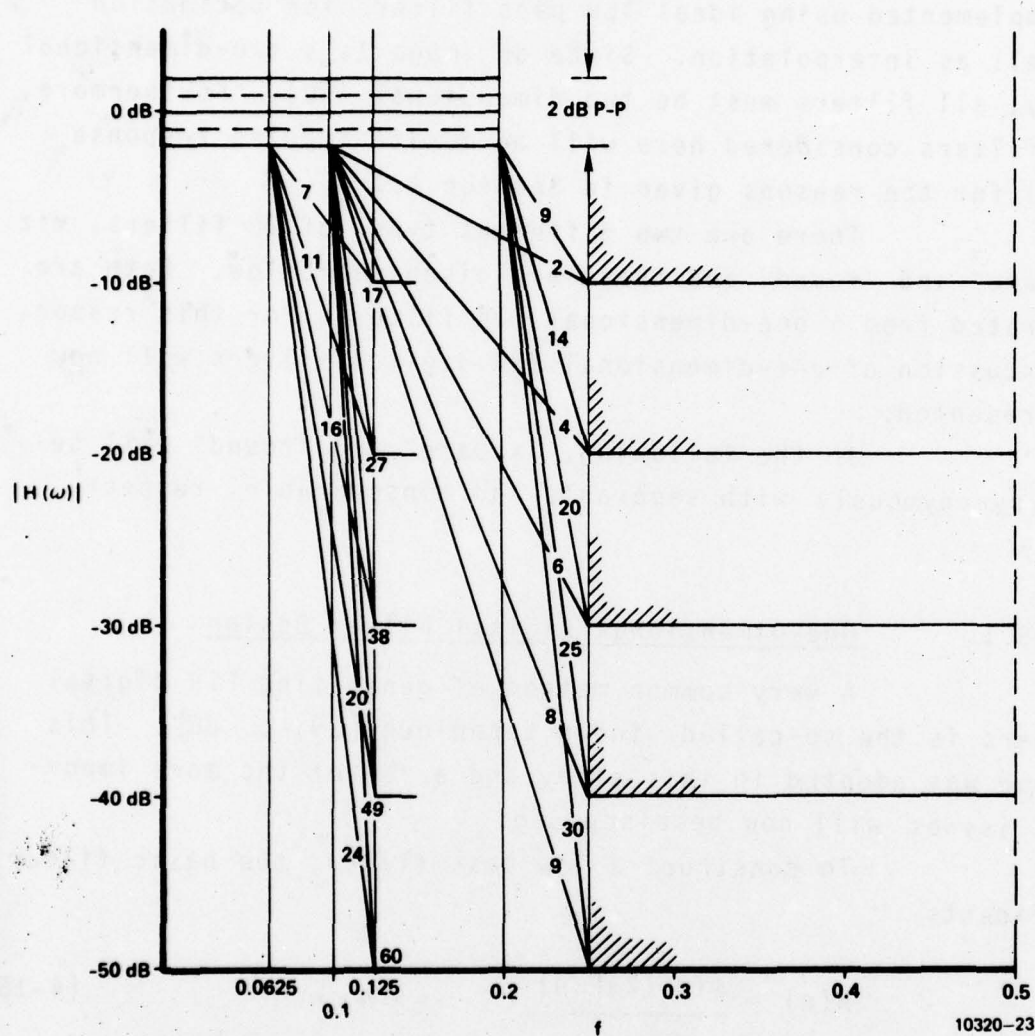


Figure 4.1.4-4  
Alternative Method of Displaying  
FIR Filter Trade-Offs

#### 4.1.5 Two-Dimensional Filters

As discussed in Section 4.1.3, the D/I process is implemented using ideal low pass filters for decimation as well as interpolation. Since an image is a two-dimensional array, all filters must be two dimensional (2D). Furthermore, all filters considered here will be finite impulse response (FIR) for the reasons given in Section 4.1.4.

There are two different types of 2D filters, viz., "square" and "round" and these are discussed below. Both are generated from a one-dimensional FIR filter. For this reason, a discussion of one-dimensional FIR digital filters will now be presented.

In the following, "square" and "round" will be used synonymously with separable and nonseparable, respectively.

##### 4.1.5.1 One-Dimensional Digital Filter Design

A very common method of generating FIR digital filters is the so-called window technique [29, p. 88]. This method was adopted in this study and a few of the more important issues will now be discussed.

To construct a low pass filter, the basic filter coefficients

$$h(n) = \frac{\sin(2\pi F n)}{\pi n}, \quad -\infty < n < \infty \quad (4-15)$$

are multiplied by a window  $w(n)$  for  $-(\frac{N-1}{2}) \leq n \leq (\frac{N-1}{2})$  where  $N$  is an odd number for all the filters to be considered here. The window  $w(n)$  to be used here is the Kaiser window and is given by

$$w(n) = \frac{I_0(\beta \sqrt{1 - [2n/(N-1)]^2})}{I_0(\beta)} \quad (4-16)$$

for  $-\left(\frac{N-1}{2}\right) \leq n \leq \left(\frac{N-1}{2}\right)$ . The parameter  $\beta$  is a constant that specifies a frequency response tradeoff between the peak height of the side lobe ripples (stopband attenuation) and the width of the transition band (see Figure 4.1.5.1-1);  $I_0(x)$  is the zeroth-order Bessel function. Table 4.1.5.1 presents the tradeoff between the transition bandwidth and stopband attenuation as  $\beta$  is varied.

TABLE 4.1.5.1

$\beta$	D	Stopband ripple (dB)
2.120	1.5	-30
3.384	2.23	-40
4.538	2.93	-50
5.658	3.62	-60
6.764	4.32	-70
7.865	5.0	-80
8.960	5.7	-90
10.056	6.4	-100

Transition bandwidth is  $D/(N-1)$ .

The filters designed by this technique have two properties worth noting: they have linear phase, which is extremely important for image processing, and the space (time) response  $h(n)$   $w(n)$  is symmetric about  $n = 0$ .

An example of a filter designed with this technique is shown in Figure 4.1.5.1-2, which gives the frequency response in dB. The parameters for this filter are indicated on the plot: the order of the filter (number of points) is 31, the Kaiser parameter  $\beta$  is 3.0, and the bandwidth  $\rho$  is 0.42.



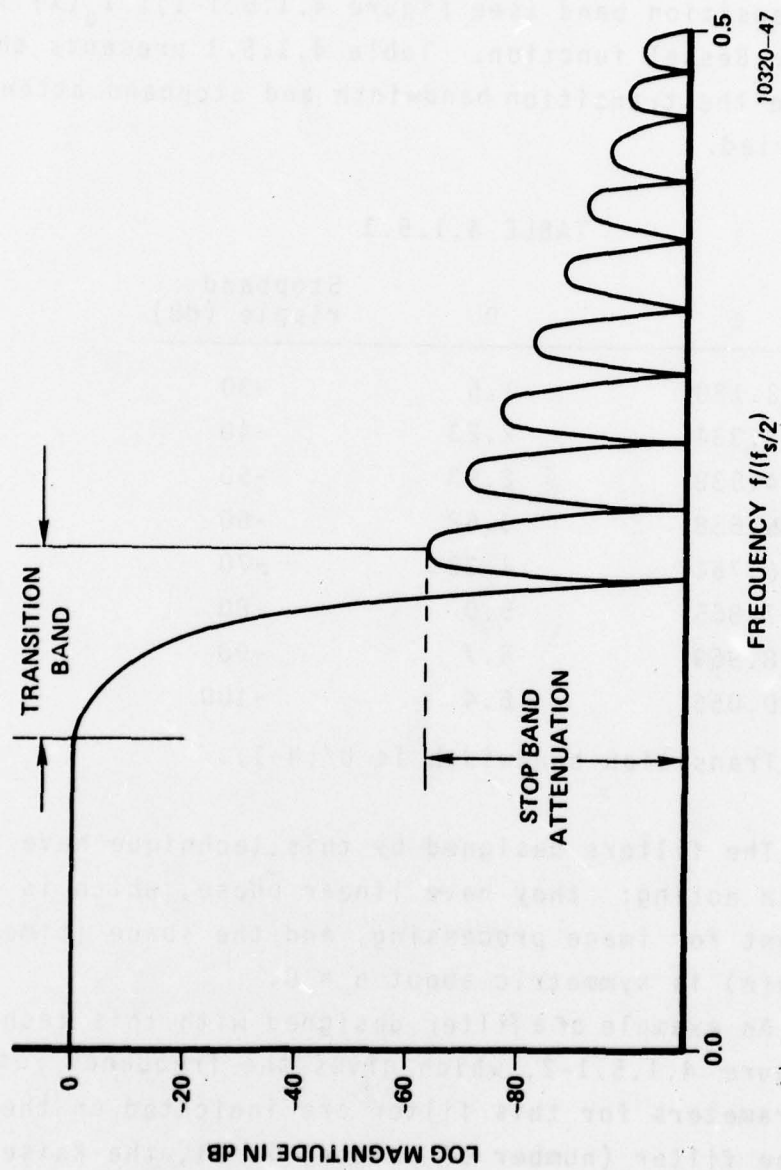


Figure 4.1.5.1-1  
Definition of Filter Parameters

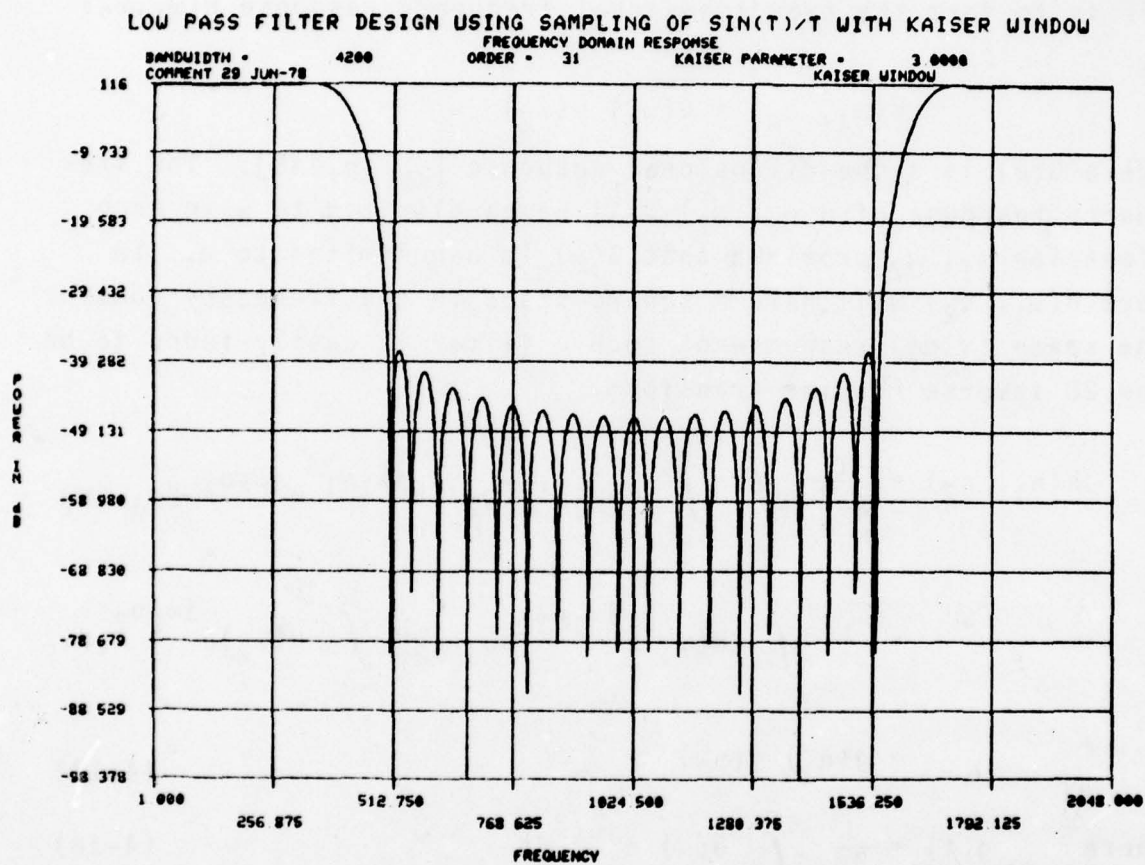


Figure 4.1.5.1-2  
 One-Dimensional LP Filter

#### 4.1.5.2

#### Two-Dimensional Digital Filter Design

The one-dimensional FIR filters described above were used to construct two-dimensional filters. There are two filter classes of interest: the "square" filter and the "round" filter.

The simplest way to generate a two-dimensional LPF is to form the two-dimensional frequency response  $H(\omega_1, \omega_2)$  as

$$H(\omega_1, \omega_2) = G(\omega_1) G(\omega_2)$$

where  $G(\omega)$  is a one-dimensional response [29, p.445]. The frequency response of  $H(\omega_1, \omega_2)$  will be bandlimited to  $\Omega$  in each dimension  $\omega_1, \omega_2$  provided that  $G(\omega)$  is bandlimited to  $\Omega$ . In fact  $H(\omega_1, \omega_2)$  will have a square shape in the frequency domain. The space (time) response of such a filter is easily found to be the 2D inverse Fourier transform.

$$\begin{aligned} h(n_1, n_2) &= \frac{1}{4\pi^2} \int_{-\Omega}^{\Omega} \int_{-\Omega}^{\Omega} H(\omega_1) H(\omega_2) e^{j\omega_1 n_1} e^{j\omega_2 n_2} d\omega_1 d\omega_2 \\ &= \frac{1}{2\pi} \int_{-\Omega}^{\Omega} H(\omega_1) e^{j\omega_1 n_1} d\omega_1 \frac{1}{2\pi} \int_{-\Omega}^{\Omega} H(\omega_2) e^{j\omega_2 n_2} d\omega_2 \\ &= g(n_1) g(n_2) \end{aligned} \quad (4-17)$$

$$\text{where} \quad g(t) = \frac{1}{2\pi} \int_{-\Omega}^{\Omega} G(\omega) e^{j\omega t} dt \quad (4-18)$$

is the space (time) response of the low pass filter  $G(\omega)$ .

To implement such an approach, one merely generates the  $g(n)$  response using the window technique described above for one-dimensional filters. This is then used to generate

$$h(n_1, n_2) = g(n_1) g(n_2) \quad (4-19)$$



for 
$$-(\frac{N-1}{2}) \leq n \leq \frac{N-1}{2} . \quad (4-20)$$

Of course, such a filter has a square shaped frequency response. Figure 4.1.5.1-2 shows a 1D frequency response generated using the window technique while Figure 4.1.5.2-1 shows a perspective plot of the 2D square (nonseparable) filter response generated from the 1D filter. As can be seen from this perspective plot, the passband of such a filter is square in nature.

The response in the horizontal and diagonal direction in the frequency domain for square filters is easily calculated. Since

$$H(\omega_1, \omega_2) = G(\omega_1) G(\omega_2) \quad (4-21)$$

and for LP filters,  $G(0) = 1$ , it can be seen that the horizontal response is

$$H(\omega_1, 0) = G(\omega_1) \cdot 1 = G(\omega_1). \quad (4-22)$$

Furthermore the diagonal response is seen to be

$$\begin{aligned} H(\omega, \omega) &= G(\omega) G(\omega) \\ &= G^2(\omega) \end{aligned} \quad (4-23)$$

Hence the diagonal response in the frequency domain is down twice as many dB as the horizontal response for square filters. This can be seen by examining Figure 4.1.5.2-1.

Since imagery has no preferred direction in the frequency domain, it seems that a square filter would not be desirable since its response is not the same in all directions. For this reason, another type of 2D filter was designed and implemented. This is the so-called round filter, developed by McClellan [34]. Basically, this approach transforms a 1D FIR filter into the 2D frequency domain using the transformation

$$\cos \omega = \frac{1}{2} [\cos \omega_1 + \cos \omega_2 + \cos \omega_1 \cos \omega_2 - 1] \quad (4-24)$$

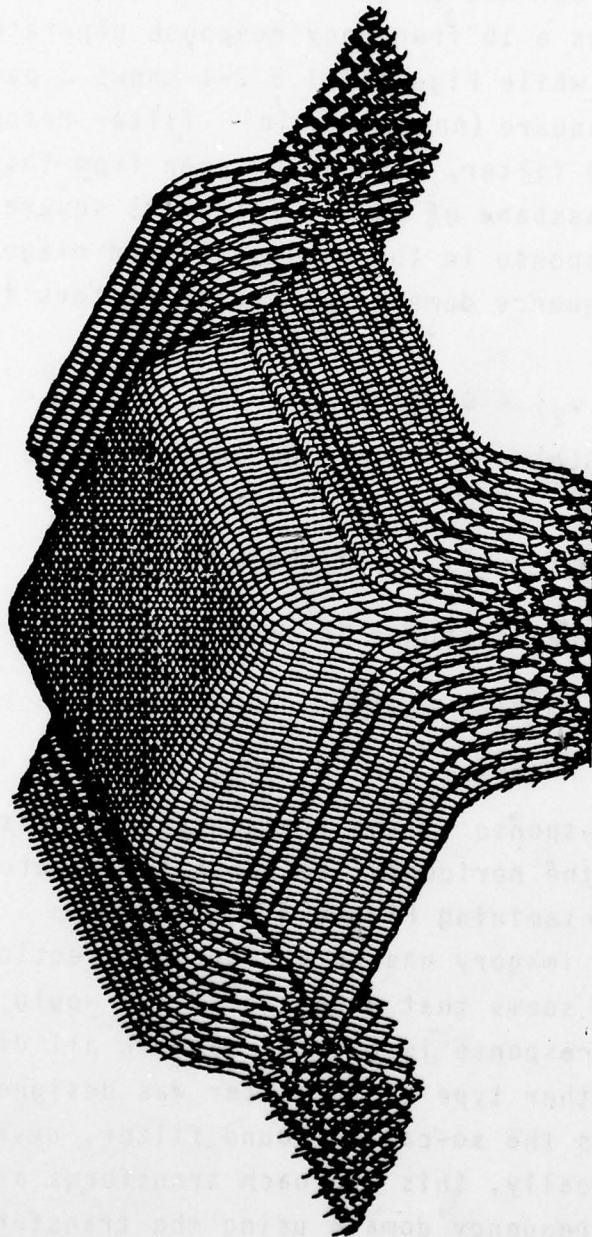


Figure 4.1.5.2-1  
Two-Dimensional Square Filter, Generated  
Using Filter of Figure 4.1.5.1-2

where  $\omega$  is the radian frequency of the 1D filter while  $(\omega_1, \omega_2)$  are the radian frequency variables of the 2D frequency domain. This transformation generates a nearly circular frequency response. The details of this transformation are given in [29, p. 472]. An example of the 2D frequency response is given in Figure 4.1.5.2-2 which was generated from the 1D filter of Figure 4.1.5.1-2. Note the circular passband response.

These 2D filters were applied to imagery in the space domain using 2D convolution. Since the response of these filters is symmetric about the center element, 2D convolution is equivalent to multiplying the 2D response  $h(n_1, n_2)$  times each pixel value and summing. This sum is the convolution. This process is repeated for each pixel value in the image. When interpolation is performed, the image is considered to consist of the decimated values along with the added zeros as discussed in Section 4.1.3.

#### 4.1.6      Two-Dimensional Decimation/Interpolation Experiments

In order to evaluate the performance of the decimation/interpolation (D/I) filters described in Sections 4.1.3 and 4.1.5, computer simulations were used. Many 2D filters, both square and round, were applied to the original image of Figure 4.1.6 (A). The original is a composite image consisting of alpha-numeric characters, an aerial view, and a frequency wedge. The frequency wedge is very useful in showing the cut-off response of the low pass filters used, as well as showing aliasing effects.

The filters were applied in the space domain by using 2D convolution. Filters of three sizes were applied: 31 x 31, 15 x 15, 9 x 9. Furthermore, the Kaiser parameter  $\beta$



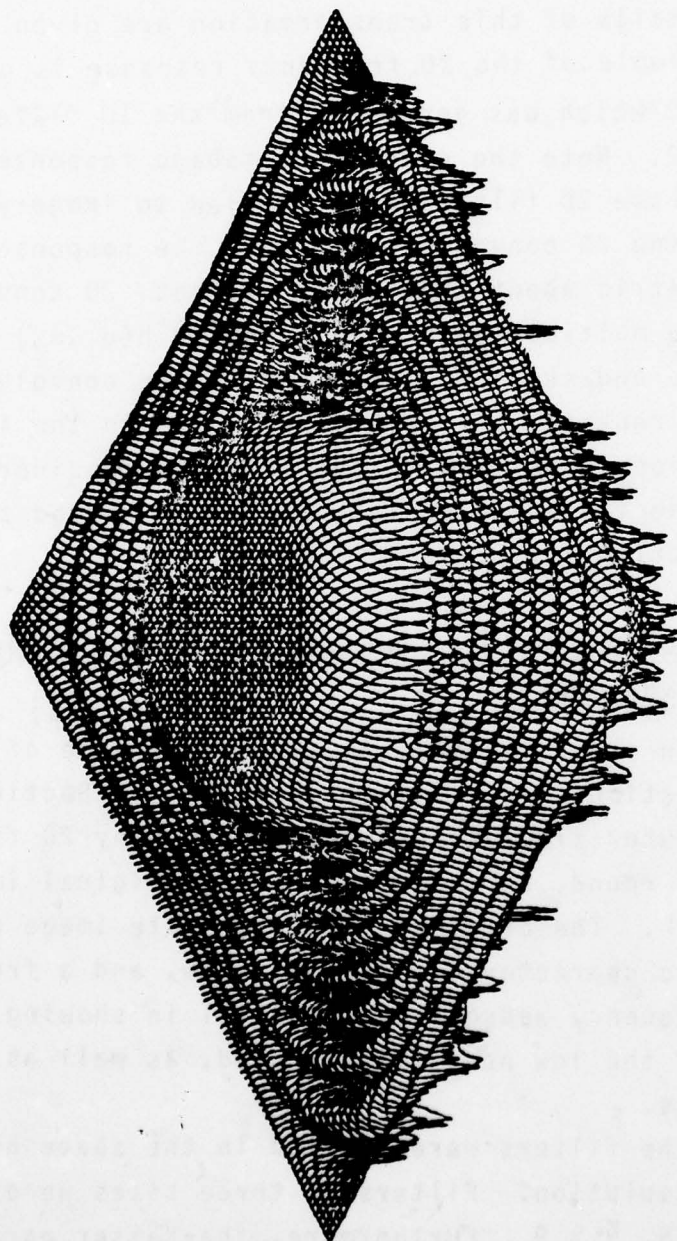


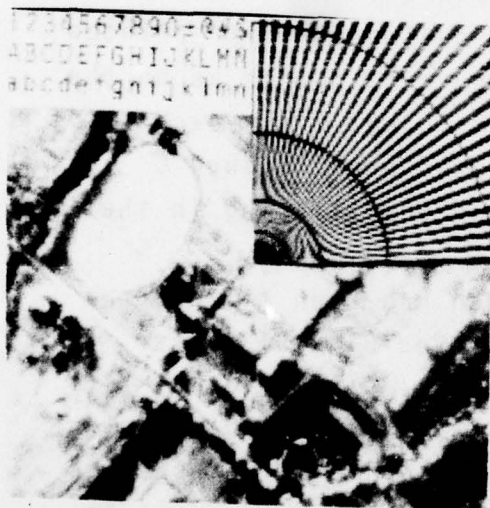
Figure 4.1.5.2-2  
Two-Dimensional Round Filter, Generated  
Using Filter of Figure 4.1.5.1-2



(A) Original



(B) 2:1 Decimate Only



(C) 2:1 D/I Only



(D) LPF only - Round  
31x31  $\beta=1.0$   $\rho=0.46$

Figure 4.1.6  
2:1 Decimation/Interpolation Computer Simulation Results

was varied from 0.0 to 3.0 giving a stopband attenuation of between -20 dB and -36 dB. The bandwidth of the filters was also varied where  $0 \leq \rho \leq 1.0$ : when  $\rho = 1.0$ , the bandwidth of the filter is  $f_s/2$ , i.e., the maximum frequency. As described in Section 4.1.5, a 2D filter may be generated from a 1D filter. Appendix I gives all the 1D filters used here. Both square and round filters were applied to the original image. In all cases, the same filter used for decimation was also used for interpolation, unless otherwise noted.

Since filtering is implemented using 2D convolution, the boundary points will not have as many points for the convolution calculation: hence these points are not reproduced and the filtered image will always be smaller than the original.

#### 4.1.6.1 Two-to-One Decimation/Interpolation

Figure 4.1.6. (B) through (P) shows various images for the 2-to-1 (2:1) decimation/interpolation (D/I) experiments. Figure B shows a typical decimated image printed at full resolution. There are  $\frac{1}{4}$  as many data points as on the original A. If trivial interpolation is used, i.e., each point is repeated once and each line is repeated once, Figure C results. Note the blocky appearance and the aliasing in the frequency wedge.

Figure D is an image which has been filtered only--no decimation/interpolation. A round  $31 \times 31$  filter, with Kaiser parameter  $\beta = 1.0$  and bandwidth  $\rho = 0.46$  was used. For an ideal LP filter,  $\rho$  should equal 0.5 for 2:1 D/I; however, when using filters of finite order,  $\rho$  can be adjusted to control the placement of the transition and stopband. It is felt that D more or less represents the best possible image that can result when doing 2:1 D/I, since only low pass filtering has been applied. If the D/I low pass filters are per-





(E) 2:1 D/I Round  
31x31  $\beta=1.0$   $\rho=0.46$



(F) 2:1 D/I Square  
31x31  $\beta=1.0$   $\rho=0.46$



(G) 2:1 D/I Round  
31x31  $\beta=0.0$   $\rho=0.45$



(H) 2:1 D/I Round  
31x31  $\beta=3.0$   $\rho=0.42$

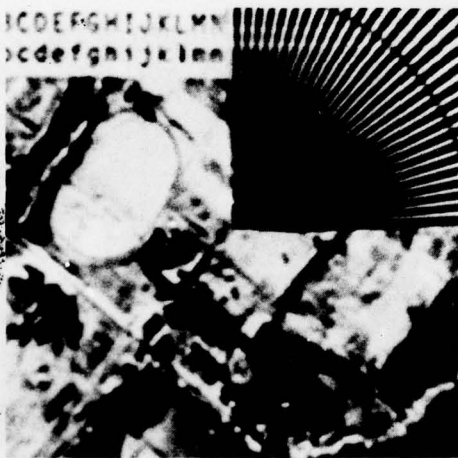
Figure 4.1.6 (continued)  
2:1 Decimation/Interpolation Computer Simulation Results



(I) 2:1 D/I Square  
31x31  $\beta=3.0$   $\rho=0.42$



(J) 2:1 D/I Round  
15x15  $\beta=1.0$   $\rho=0.4$

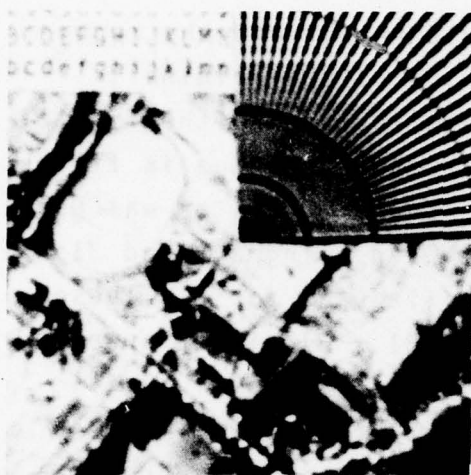


(K) 2:1 D/I Square  
15x15  $\beta=3.0$   $\rho=0.35$



(L) 2:1 D/I Round  
15x15  $\beta=0.0$   $\rho=0.41$

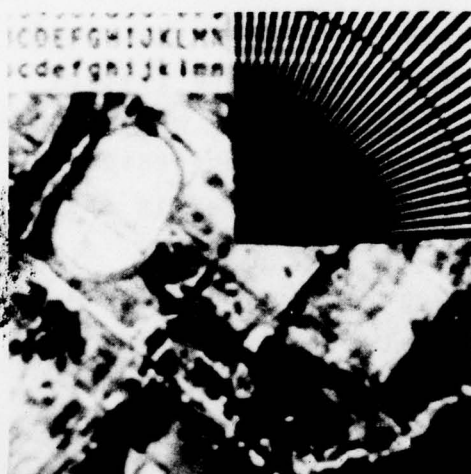
Figure 4.1.6 (continued)  
Decimation/Interpolation Computer Simulation Results



(M) 2:1 D/I Round  
9x9  $\beta=0.0$   $\rho=0.37$



(N) 2:1 D/I Round  
9x9  $\beta=1.0$   $\rho=0.355$



(O) 2:1 D/I Square  
9x9  $\beta=1.0$   $\rho=0.355$



(P) 2:1 D/I Round  
9x9  $\beta=3.0$   $\rho=0.21$

Figure 4.1.6 (continued)  
2:1 Decimation/Interpolation Computer Simulation Results



forming correctly, then no aliasing would result and the reconstructed image would be a low pass version of the original; hence, D is the best possible image (notwithstanding the finite size of the  $31 \times 31$  filter). Note the frequency wedge is filtered equally in all directions since a round 2D filter was used.

Figure E shows a 2:1 D/I when using a round  $31 \times 31$ ,  $\beta = 1.0$ ,  $\rho = 0.46$  filter, the same filter as for D. These images are essentially the same; hence a  $31 \times 31$  filter appears to be a large enough filter for 2:1 D/I.

Figure F is a square  $31 \times 31$ ,  $\beta = 1.0$ ,  $\rho = 0.46$  filter. Hence the E and F filters differ only in that one is round and the other is square. Note that the images are virtually identical - only the frequency wedge appears different with the diagonal frequencies attenuated at a higher frequency than the horizontal (or vertical) frequencies. (It should be noted that the passband of a square LP filter is  $\sqrt{2}$  greater in the diagonal direction than in the horizontal or vertical directions).

In these experiments it was found that the round filter was virtually indistinguishable from the square filter, all other parameters being equal: only the frequency wedge shows any difference with higher diagonal frequencies being passed. This is an important observation since a square filter is easier to implement as will be shown in Section 4.1.7

Figures G through I show a variety of  $31 \times 31$  filters and these demonstrate the fact that slight changes in the bandwidth and attenuation of the filters do not significantly affect the D/I performance for a  $31 \times 31$  filter.

Figures J through L show a variety of 15 x 15 filters. They appear to be somewhat degraded from the 31 x 31 filters.

Figures M through P show a variety of 9 x 9 filters. Note that if the bandwidth  $\rho$  of the filter is too small, the interpolated image is blurred (see Figure P). Otherwise, a 9 x 9 filter generates an interpolated image somewhat lower in quality than the larger filters.

From these 2:1 D/I experiments, the following conclusions can be drawn:

- a) A 31 x 31 filter is more than adequate for 2:1 D/I
- b) A square filter can be used
- c) A 9 x 9 square  $\beta = 1.0$   $\rho = 0.355$  filter yields good performance, slightly degraded from the optimum LP filter.

#### 4.1.6.2 Four-to-One Decimation/Interpolation

Figure 4.1.6.2 B through AA shows a wide variety of 4:1 D/I experiments. Figure A is a 4:1 decimated image only; it contains 1/16th as many data points as the original of Figure 4.1.6 A. If trivial interpolation is used, i.e., each point is repeated 3 additional times and each line is repeated 3 additional times, figure B results. Note the extreme blockiness and the severe aliasing in the frequency wedge.

Figure C is an image which has been filtered only--no D/I. A 31 x 31 filter was used and represents the best possible D/I image possible, notwithstanding the finite size of the 31 x 31 filter. (For 4:1 D/I  $\rho$  should equal 0.25 when using an ideal LPF: it may differ from this for a finite order filter.) It is felt that C represents the best possible

image that can result when doing 4:1 D/I since only low pass filtering has been applied. The remainder of the images of Figure 4.1.6.2 involve both decimation and interpolation.

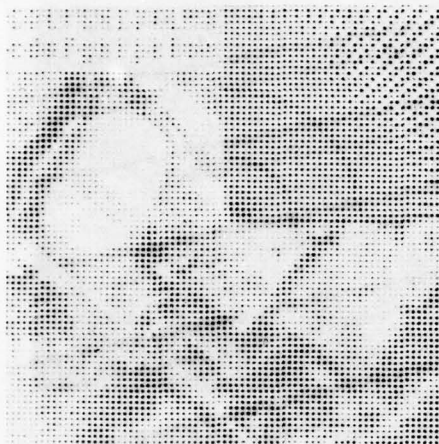
Figures D through K give a variety of 31 x 31 filters varying the same parameters as for the 2:1 D/I experiments above. It will be noted that  $\rho = 0.25$  and  $\beta = 1.0, 3.0$  give the best results. A smaller  $\rho$  or smaller  $\beta$  gives a blurred image; a larger  $\rho$  (figure K) gives a blocky image due to the fact that aliasing is occurring.

It will be noted that the round and square filters are virtually indistinguishable from each other: only the frequency wedge is filtered differently with the square filter allowing higher diagonal frequencies to pass. This result is true for all size filters and is important since square filters are easier to implement than around filters (see Section 4.1.7).

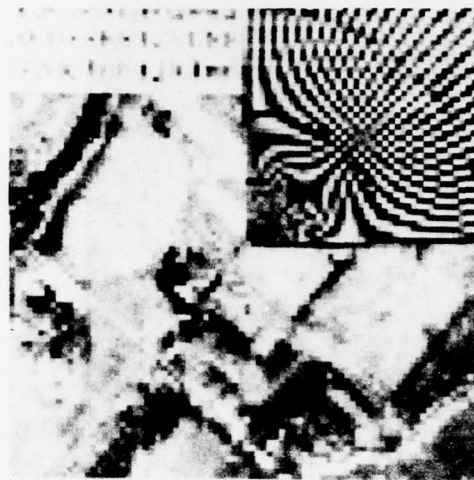
Figures L through P show the 15 x 15 filtered images. Note that  $\rho = 0.25$  and  $\beta = 1.0, 3.0$  give the best results. If  $\rho$  is smaller the image appears blurry (filter rejecting too much information); if  $\beta = 0$ , the image appears blocky (filter allowing aliasing).

Figures Q through X show the 9 x 9 filter images. The best image appears to be the  $\beta = 3.0$  with a low bandwidth of  $\rho = 0.01$ . This is because a 9 x 9 filter has a very wide transition bandwidth and if  $\rho$  is any larger, aliasing will occur. This is verified by the reconstructed images. If  $\rho$  is small and  $\beta = 1.0$ , then aliasing will dominate since a  $\beta = 1.0$  filter gives only about -26 dB stopband attenuation as opposed to -36 dB for  $\beta = 3.0$ . This is seen in Figure W, X.

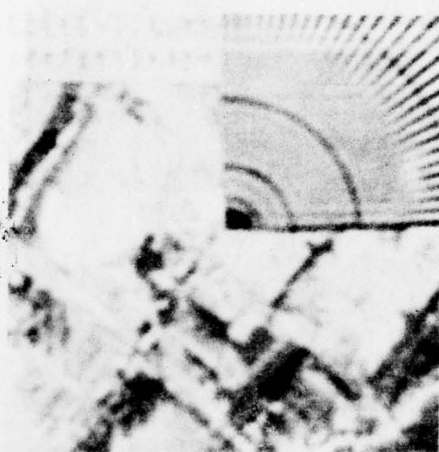




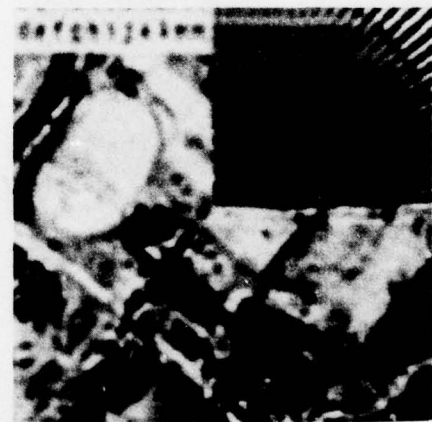
(A) 4:1 Decimate Only



(B) 4:1 D/I Only



(C) LPF Only Round  
31x31  $\beta=1.0$   $\rho=0.25$



(D) 4:1 D/I Round  
31x31  $\beta=1.0$   $\rho=0.25$

Figure 4.1.6.2  
4:1 Decimation/Interpolation Computer Simulation Results

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ARRAY SCANNING FOR IMAGE TRANSMISSION.(U)  
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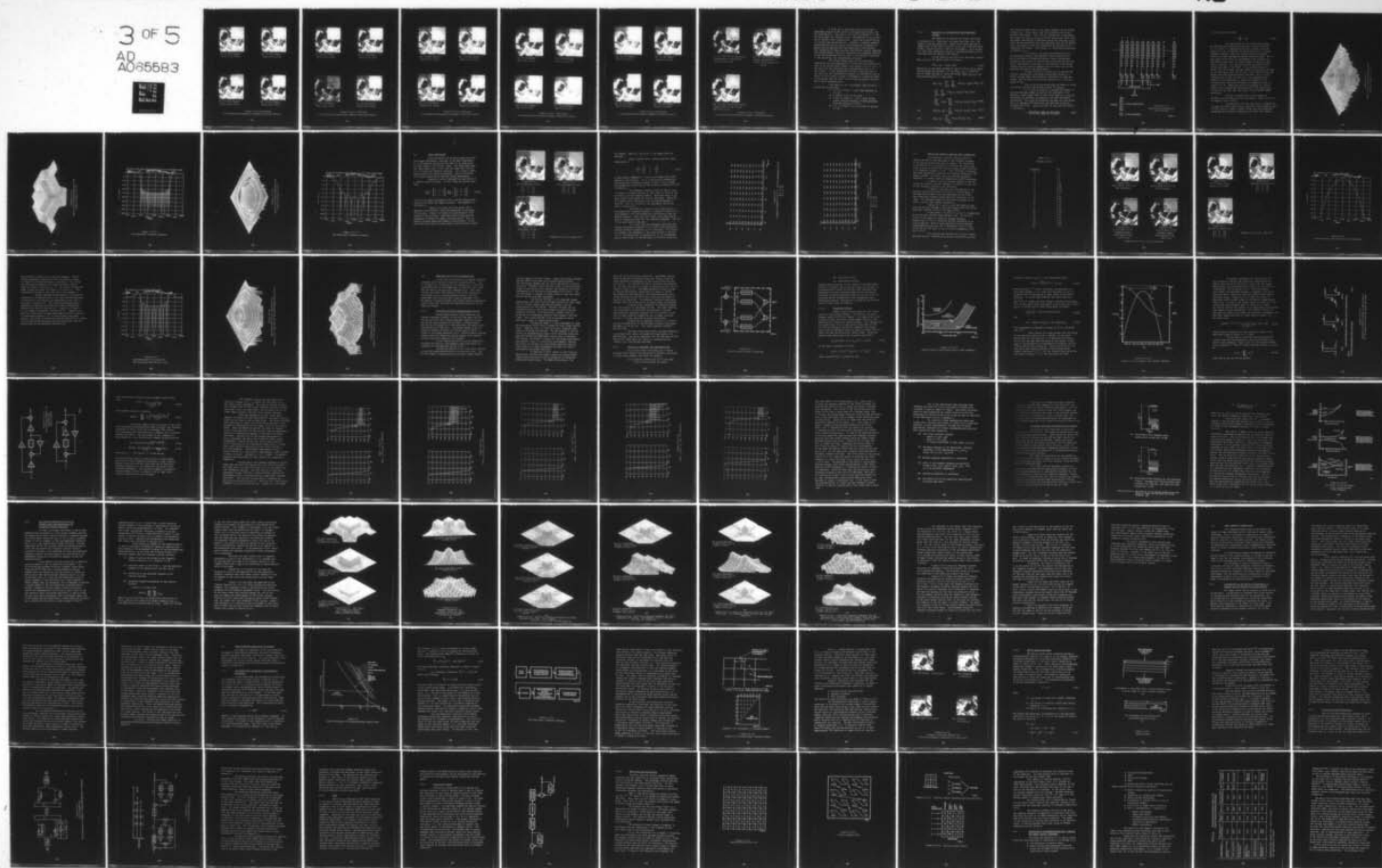
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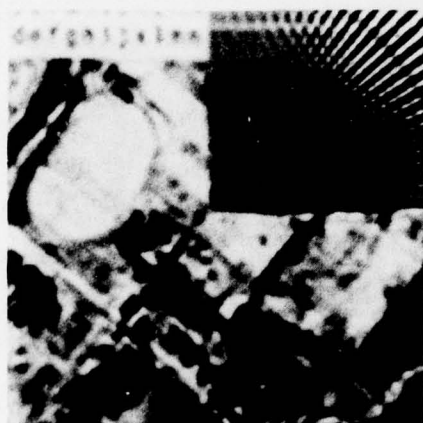




(E) 4:1 D/I Square  
31x31  $\beta=1.0$   $\rho=0.25$



(F) 4:1 D/I Round  
31x31  $\beta=3.0$   $\rho=0.25$



(G) 4:1 D/I Square  
31x31  $\beta=3.0$   $\rho=0.25$



(H) 4:1 D/I Round  
31x31  $\beta=0.0$   $\rho=0.25$

Figure 4.1.6.2 (continued)  
4:1 Decimation/Interpolation Computer Simulation Results

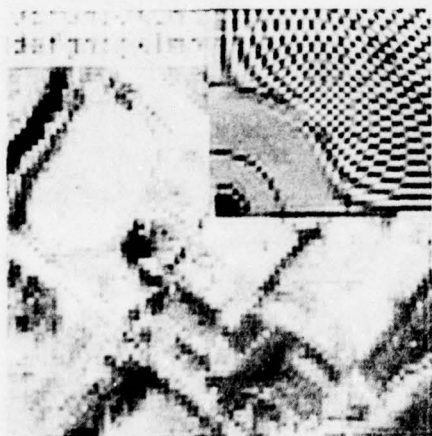




(I) 4:1 D/I Round  
31x31  $\beta=1.0$   $\rho=0.21$



(J) 4:1 D/I Round  
31x31  $\beta=3.0$   $\rho=0.17$

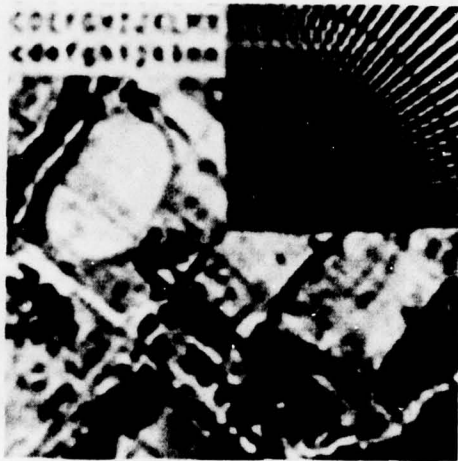


(K) 4:1 D/I Round  
31x31  $\beta=3.0$   $\rho=0.42$



(L) 4:1 D/I Round  
15x15  $\beta=1.0$   $\rho=0.25$

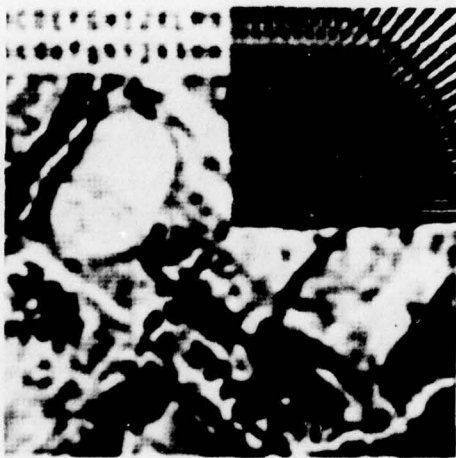
Figure 4.1.6.2 (continued)  
4:1 Decimation/Interpolation Computer Simulation Results



(M) 4:1 D/I Square  
15x15  $\beta=1.0$   $\rho=0.25$



(N) 4:1 D/I Square  
15x15  $\beta=3.0$   $\rho=0.25$



(O) 4:1 D/I Round  
15x15  $\beta=0.0$   $\rho=0.25$

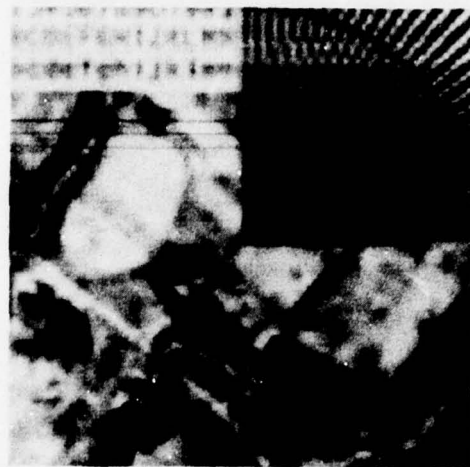


(P) 4:1 D/I Round  
15x15  $\beta=1.0$   $\rho=0.14$

Figure 4.1.6.2 (continued)  
4:1 Decimation/Interpolation Computer Simulation Results



(Q) 4:1 D/I Round  
9x9  $\beta=3.0$   $\rho=0.01$



(R) 4:1 D/I Round  
9x9  $\beta=3.0$   $\rho=0.25$



(S) 4:1 D/I Square  
9x9  $\beta=3.0$   $\rho=0.25$



(T) 4:1 D/I Round  
9x9  $\beta=1.0$   $\rho=0.25$

Figure 4.1.6.2 (continued)  
4:1 Decimation/Interpolation Computer Simulation Results





(U) 4:1 D/I Square  
9x9  $\beta=1.0$   $\rho=0.25$



(V) 4:1 D/I Round  
9x9  $\beta=0.0$   $\rho=0.25$

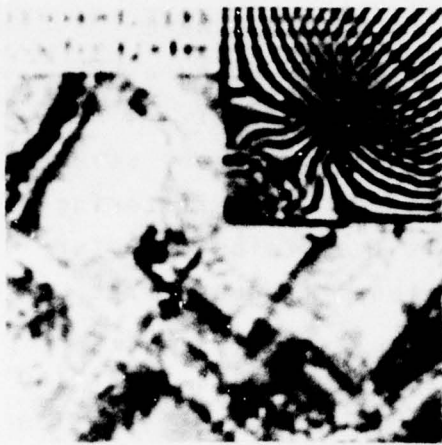


(W) 4:1 D/I Round  
9x9  $\beta=1.0$   $\rho=0.05$

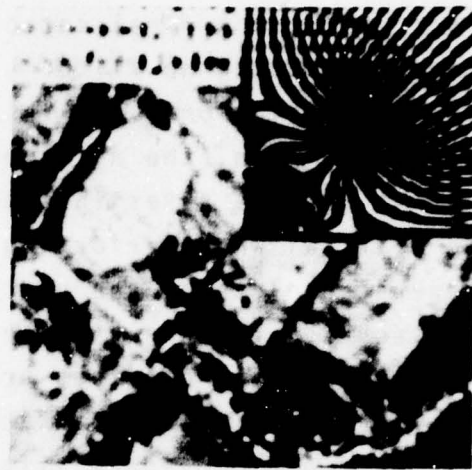


(X) 4:1 D/I Square  
9x9  $\beta=1.0$   $\rho=0.05$

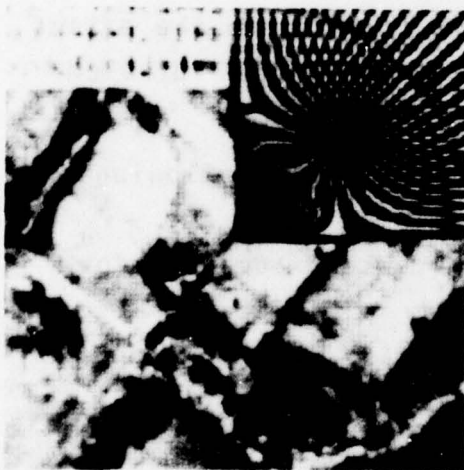
Figure 4.1.6.2 (continued)  
4:1 Decimation/Interpolation Computer Simulation Results



(Y) 4:1 D/I Trivial Decimation  
Interpolation Filter Round  
31x31  $\beta=1.0$   $\rho=0.25$



(Z) 4:1 D/I Trivial Decimation  
Interpolation Filter Round 15x15  $\beta=1.0$   
 $\rho=0.25$



(AA) 4:1 D/I Trivial Decimation  
Interpolation Filter Square  
9x9  $\beta=1.0$   $\rho=0.25$

Figure 4.1.6.2 (continued)  
4:1 Decimation/Interpolation Computer Simulation Results

A final set of experiments for 4:1 D/I were performed using no filter prior to decimation but using a LPF for interpolation. The results are shown in Figures Y, Z, AA. As can be seen, the aliasing error from sampling is very severe and it remains after interpolating. Hence the LP prefiltering is critical when performing D/I. Appendix I contains computer-generated plots of the various one-dimensional frequency responses used to generate the two-dimensional filter designs. This collection of plots provides quantitative results regarding the in-band ripple, transition bandwidth, and stopband rejection of the two-dimensional filters used in the computer simulation of the decimation and interpolation processes.

One unanticipated result of the simulation experiments was the appearance of a fine grid-like structure in some of the reconstructed images (see Figure 4.1.6.2 V, W, X, Z, AA) which had been decimated and interpolated using  $9 \times 9$  FIR filters. It was determined that this effect was due to the nature of the interpolation process; an explanation and analysis of this effect is given in Appendix J. Since the effect is noticeable only for small filter sizes it poses no problem for an ASITS implementation.

From these 4:1 D/I experiments, the following conclusions can be drawn:

- a) A  $31 \times 31$  filter is more than adequate for 4:1 D/I
- b) A square filter can be used
- c) A  $15 \times 15$  square  $\beta = 1.0$   $\rho = 0.25$  filter yields good performance, slightly degraded from the optimum LP filter.
- d) A  $9 \times 9$  filter is not sufficient to perform 4:1 D/I.



## 4.1.7

Separable vs. Nonseparable Two-Dimensional  
Filters

The design of 2D digital FIR filters was given in Section 4.1.5. There were two types discussed, viz., the round LPF and the square LPF. These filters were applied to imagery and, as was seen in Section 4.1.6, the round and square filters yielded virtually identical performance. This section will show that substantial savings in hardware/software can be realized using the square filter design.

As was shown in section 4.1.5, the space response  $\bar{h}(n_1, n_2)$  of a 2D square filter is given as

$$\bar{h}(n_1, n_2) = h(n_1) h(n_2) \quad (4-25)$$

where  $h(n)$  is the space response of the 1D filter used to generate the square 2D response. If this  $\bar{h}(n_1, n_2)$  is convolved with an image  $x(p_1, p_2)$  to yield the output image  $y(m_1, m_2)$ , the convolution equation is applicable:

$$\begin{aligned} y(m_1, m_2) &= \sum_{p_1=0}^{N-1} \sum_{p_2=0}^{N-1} x(m_1-p_1, m_2-p_2) \bar{h}(p_1, p_2) \\ &= \sum_{p_1=0}^{N-1} \sum_{p_2=0}^{N-1} x(m_1-p_1, m_2-p_2) h(p_1) h(p_2) \\ &= \sum_{p_1=0}^{N-1} h(p_1) \sum_{p_2=0}^{N-1} x(m_1-p_1, m_2-p_2) h(p_2) \quad (4-26) \end{aligned}$$

Let 
$$q(m_1-p_1, m_2) = \sum_{p_2=0}^{N-1} x(m_1-p_1, m_2-p_2) h(p_2) \quad (4-27)$$

Then 
$$y(m_1, m_2) = \sum_{p_1=0}^{N-1} h(p_1) q(m_1-p_1, m_2). \quad (4-28)$$

This equation states that if the space response can be written as  $\bar{h}(n_1, n_2) = h(n_1) h(n_2)$ , then 2D convolution can be performed in two steps: first vertical convolution is performed with  $h(p_2)$  and  $x(m_1-p_1, m_2-p_2)$  to yield  $q(m_1-p_1, m_2)$  after which horizontal convolution is performed using  $h(p_1)$  and  $q(m_1-p_1, m_2)$ . Since the 2D convolution has been separated into two 1D convolutions, the square filter will henceforth be referred to as a "separable" 2D filter. This process is illustrated graphically in Figure 4.1.7-1.

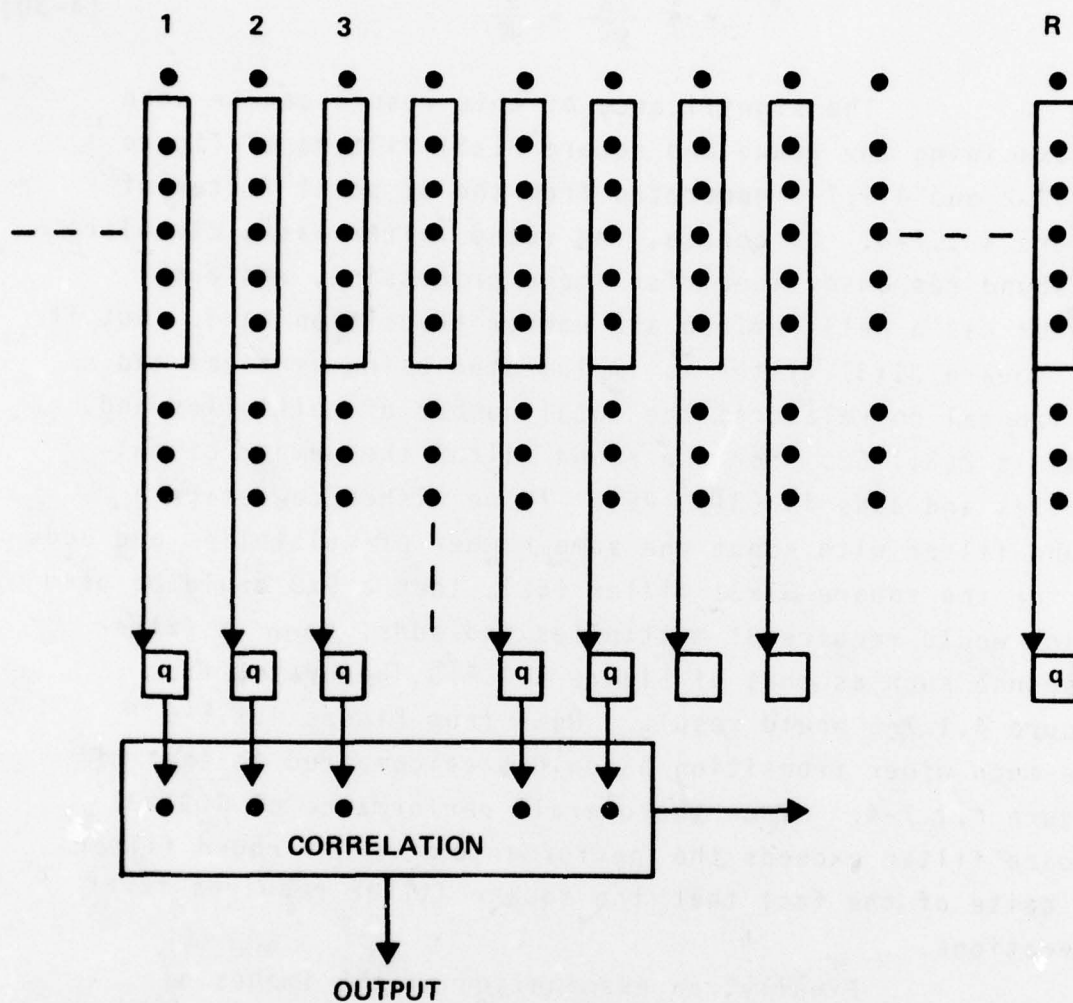
It is easy to see that each output point of a 2D convolution using a  $N \times N$  separable filter requires an  $N$  point correlation followed by another  $N$  point correlation. (Actually, this result is not strictly true since  $(N-1)/2$  vertical convolutions on the ends of the data must first be performed, but this will be ignored since it occurs only at the beginning and end of the convolution.)

If one utilizes a 1D array scan to perform the vertical convolution, with these convolutions stored in a buffer storage where the horizontal convolution is then performed, then an array of size  $N$  can perform an  $N \times N$  2D separable filter.

This is to be contrasted with the method of using a 2D array scan which would have to have  $N^2$  elements.

It is also instructive to compare the number of multiply-and-adds needed to perform a 2D separable filter function. Each output point of the filter requires a vertical correlation of  $N$  points, followed by an  $N$  point horizontal correlation. This give  $2N$  multiplies and adds for each output point as opposed to  $N^2$  multiplies and adds when using non-separable 2D filter. Hence the ratio of required calculations  $r$ , defined as

$$r = \frac{\text{multiplies, adds for 2D square}}{\text{multiplies, adds for 2D round}} \quad (4-29)$$



LEGEND:



5 POINT CORRELATION



STORAGE REGISTER

Figure 4.1.7-1  
Separable Two-Dimensional  
FIR Filtering

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is, as we have just shown

$$r = \frac{2N}{N^2} = \frac{2}{N} . \quad (4-30)$$

The significance of this result can be seen by examining the round and square 31x31 filters of Figure 4.1.7-2 and 4.1.7-3 generated from the 31 point filter of Figure 4.1.7-4. Of course, the round filter has a circular passband response, ideal for image processing, and each filter has a well defined and narrow transition band. But if the square 31x31 filter is implemented using vertical and horizontal correlators, the total number of multiplies and adds is  $2(31)=62$ . For the round filter the number of multiplies and adds is  $(31)^2=961$ . If we wished to design a round filter with about the same number of multiplies and adds as for the square 31x31 filter (62), then a 9x9 could be used which would require 81 multiplies and adds. Then a filter response such as that of Figure 4.1.7-5, generated from Figure 4.1.7-6 would result. Note from Figure 4.1.7-6 the much wider transition bandwidth as compared to that of Figure 4.1.7-4. Hence the overall performance of a 31x31 square filter exceeds the performance of a 9x9 round filter in spite of the fact that the square filter requires fewer operations.

Finally, an examination of the images of Section 4.1.6 will bear out this conclusion: Figure 4.1.6.2-E is superior to 4.1.6.2-Q.

In conclusion, for the same level of complexity, a square 2D filter rather than a round 2D filter should be used for D/I since it yields a better filter when the transition bandwidth and stop-band attenuation are taken into account.

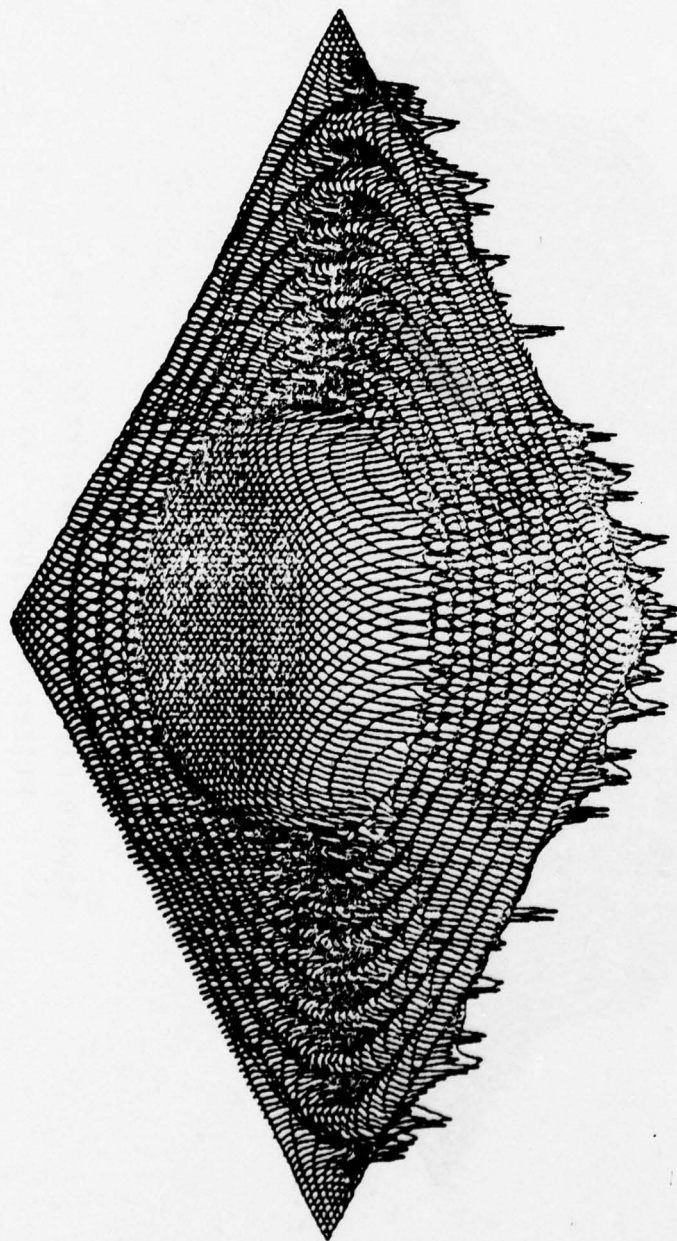


Figure 4.1.7-2  
Round Filter (31x31) Generated From  
Filter of Figure 4.1.7-4

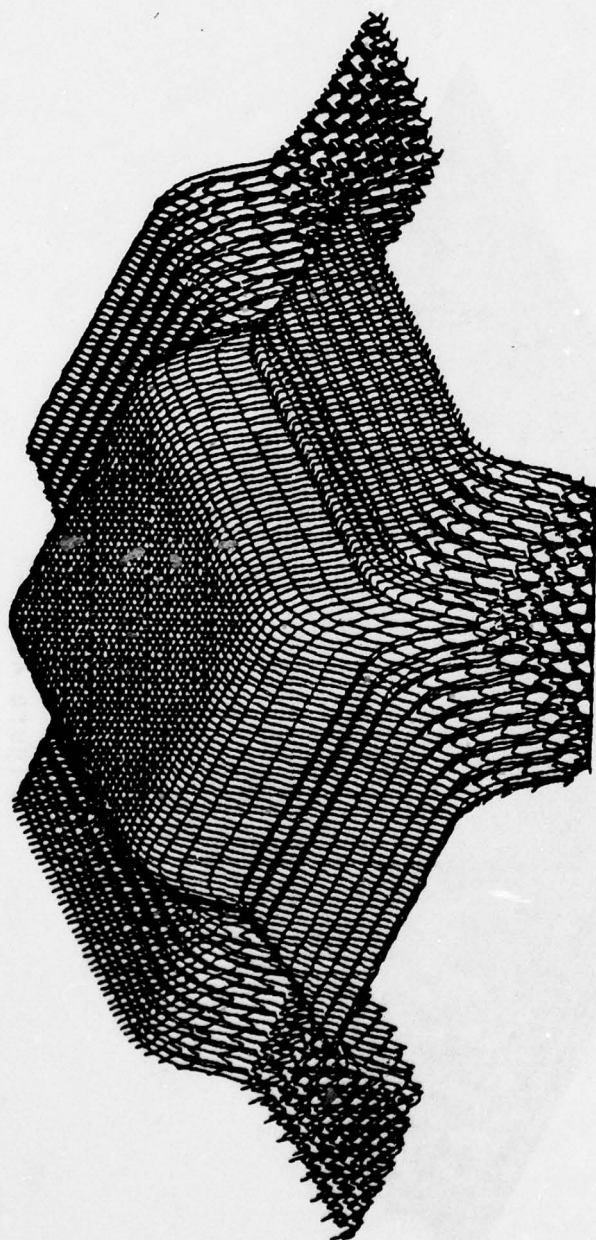


Figure 4.1.7-3  
Square Filter (31x31) Generated From  
Filter of Figure 4.1.7-4



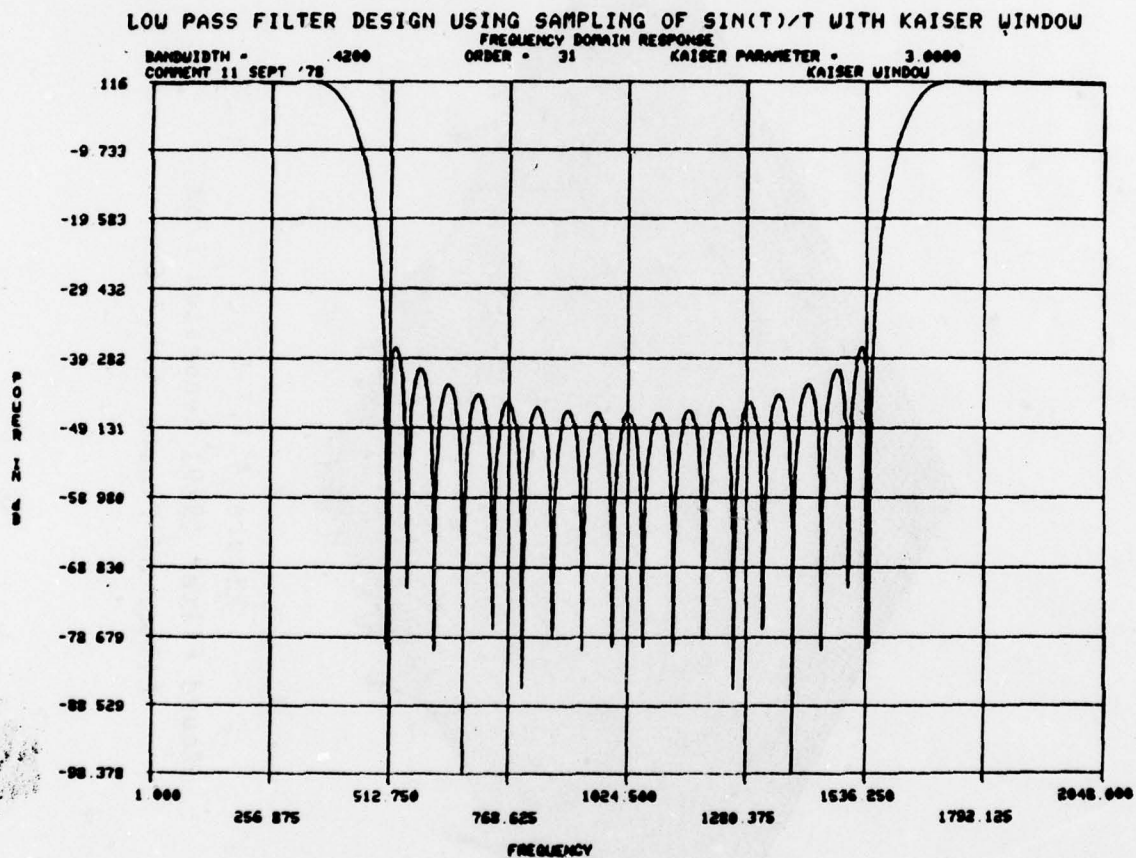


Figure 4.1.7-4  
 One-Dimensional Frequency Response

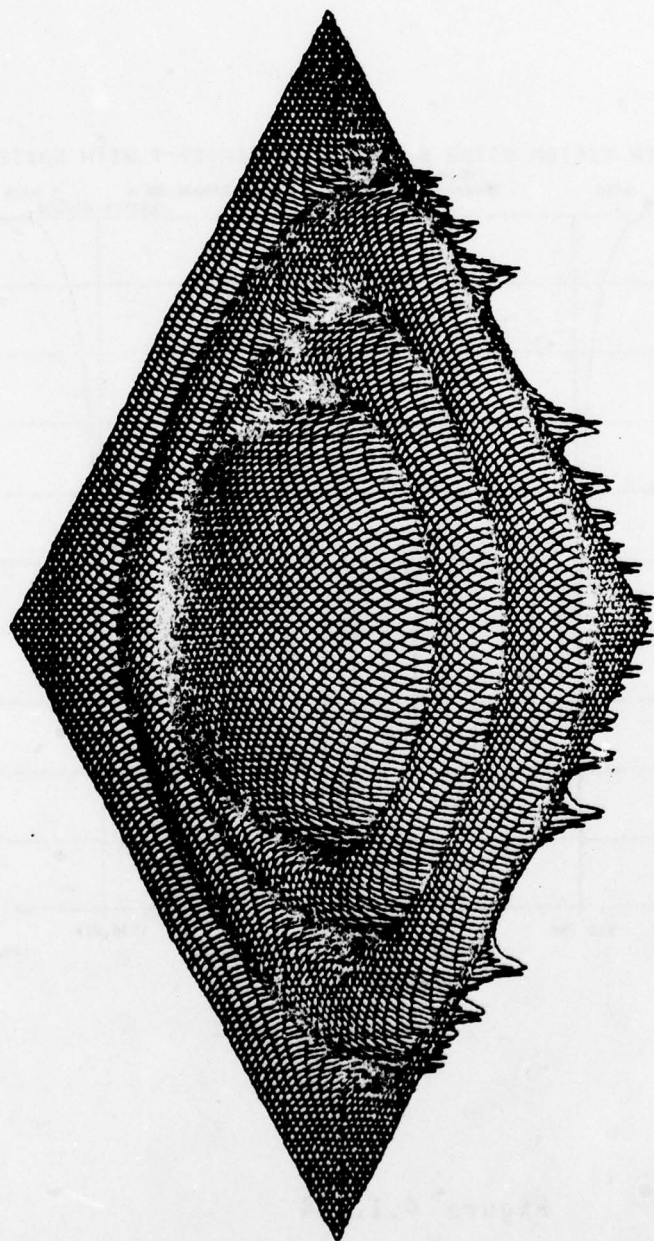


Figure 4.1.7-5  
Round Filter (9x9) Generated From  
Filter of Figure 4.1.7-6

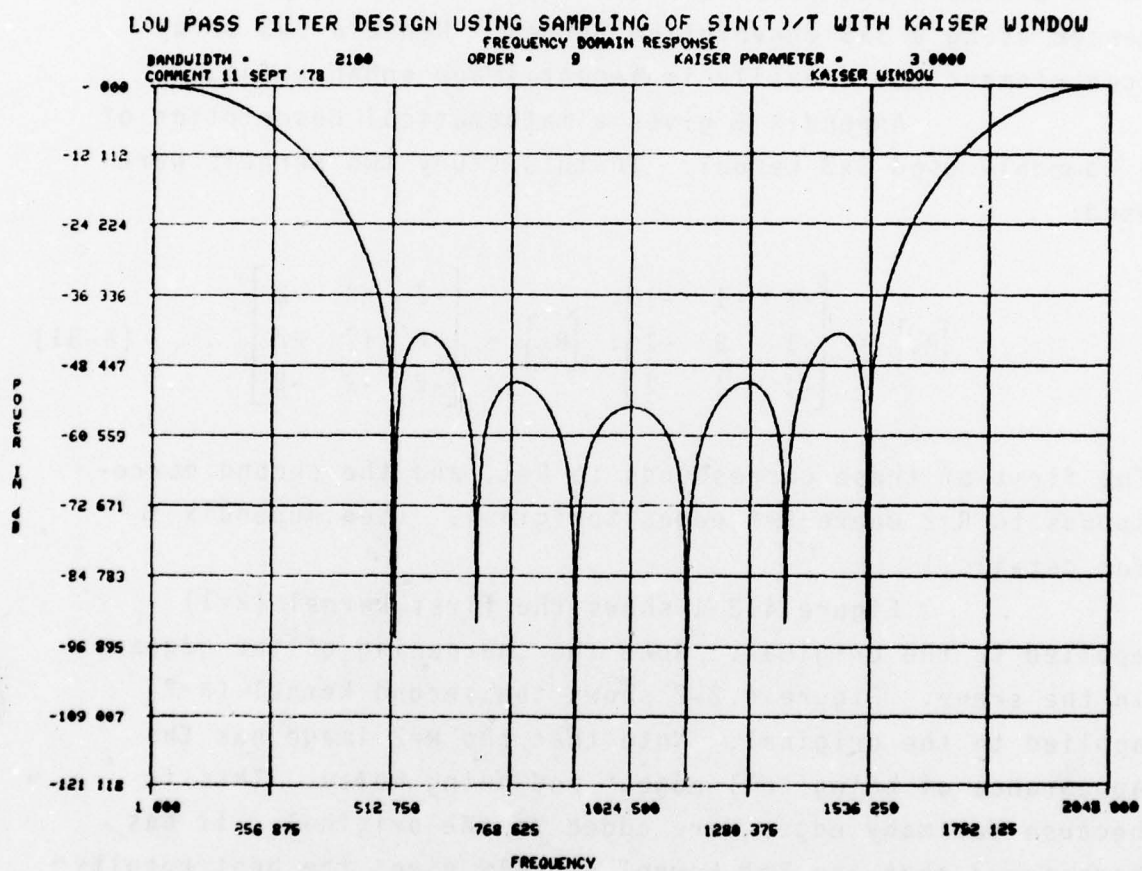


Figure 4.1.7-6  
 One-Dimensional Frequency Response



## 4.2

### IMAGE ENHANCEMENT

Array processing can be easily applied to perform image enhancement, sometimes called edge enhancement. It is performed by generating the edges of an image which are then added to the original image. The added edges make the image appear more "crisp" with the edges more visible. It is a simple but powerful technique. It is usually implemented using a 3x3 convolution kernel. Hence a 3x3 array scan element could easily implement image enhancement.

Appendix G gives a mathematical description of a commonly used 3x3 kernel. In this study two kernels were used:

$$\begin{bmatrix} R_1 \end{bmatrix} = \begin{bmatrix} -1 & -1 & -1 \\ -1 & 9 & -1 \\ -1 & -1 & -1 \end{bmatrix}, \quad \begin{bmatrix} R_2 \end{bmatrix} = \begin{bmatrix} -2 & -2 & -2 \\ -2 & 17 & -2 \\ -2 & -2 & -2 \end{bmatrix} . \quad (4-31)$$

The first of these corresponds to R=1, and the second corresponds to R=2 where R=% edges/%original. (See Appendix G for details.)

Figure 4.2-1 shows the first kernel (R=1) applied to the original. Note the sharpening of the edges in the scene. Figure 4.2-2 shows the second kernel (R=2) applied to the original. Note that the R=2 image has the appearance of being "all edges" and being noisy. This is because too many edges were added to the original. It has been found that the R=1 kernel usually gives the best results



Figure 4.2-1

R=1 Enhancement Filter

$$\begin{bmatrix} -1 & -1 & -1 \\ -1 & 9 & -1 \\ -1 & -1 & -1 \end{bmatrix}$$



Figure 4.2-2

R=2 Enhancement Filter

$$\begin{bmatrix} -2 & -2 & -2 \\ -2 & 17 & -2 \\ -2 & -2 & -2 \end{bmatrix}$$

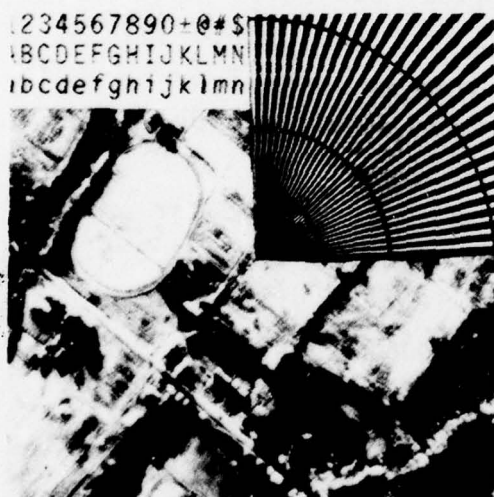


Figure 4.2-3

Enhancement Filter

$$\begin{bmatrix} 1 & -2 & 1 \\ -2 & 5 & -2 \\ 1 & -2 & 1 \end{bmatrix}$$

Figures 4.2-1, 4.2-2 and 4.2-3

for imagery. When  $R > 1$ , the noise in the image starts to dominate.

Another kernel which could be used for image enhancement is

$$[S] = \begin{bmatrix} 1 & -2 & 1 \\ -2 & 5 & -2 \\ 1 & -2 & 1 \end{bmatrix} \quad (4-32)$$

as described in Appendix H. Figure 4.2-3 shows this kernel applied to the original. It is instructive to examine the two-dimensional frequency response of  $[S]$  as well as  $[R_1]$ . The mathematical derivation of the frequency response is given in Appendix H.

The frequency response in dB for  $[S]$  is given in Figure 4.2-4 where the gain of the filter is given at the grid points of the frequency domain. Note that there is no gain along the horizontal or vertical direction. Hence  $[S]$  will enhance only frequencies that are not horizontal or vertical. This is, of course, not desirable, and can be seen in Figure 4.2-3.

The kernel  $R_1$ , however, has a more circular two-dimensional frequency response, as shown in Figure 4.2-5. (See Appendix H for the mathematical derivation of the frequency response.) Note that this kernel gives about 20 dB gain at frequencies near half the sampling frequency. The  $R=2$  filter gives about 25 dB gain.

It is seen from the frequency plots that the kernels described in Appendix G will yield good performance and will enhance frequencies in all directions. This kernel is, however, not separable in the sense defined in Section 4.1.7. The kernel  $[S]$  is separable but is not recommended due to lack of gain in the horizontal or vertical direction.



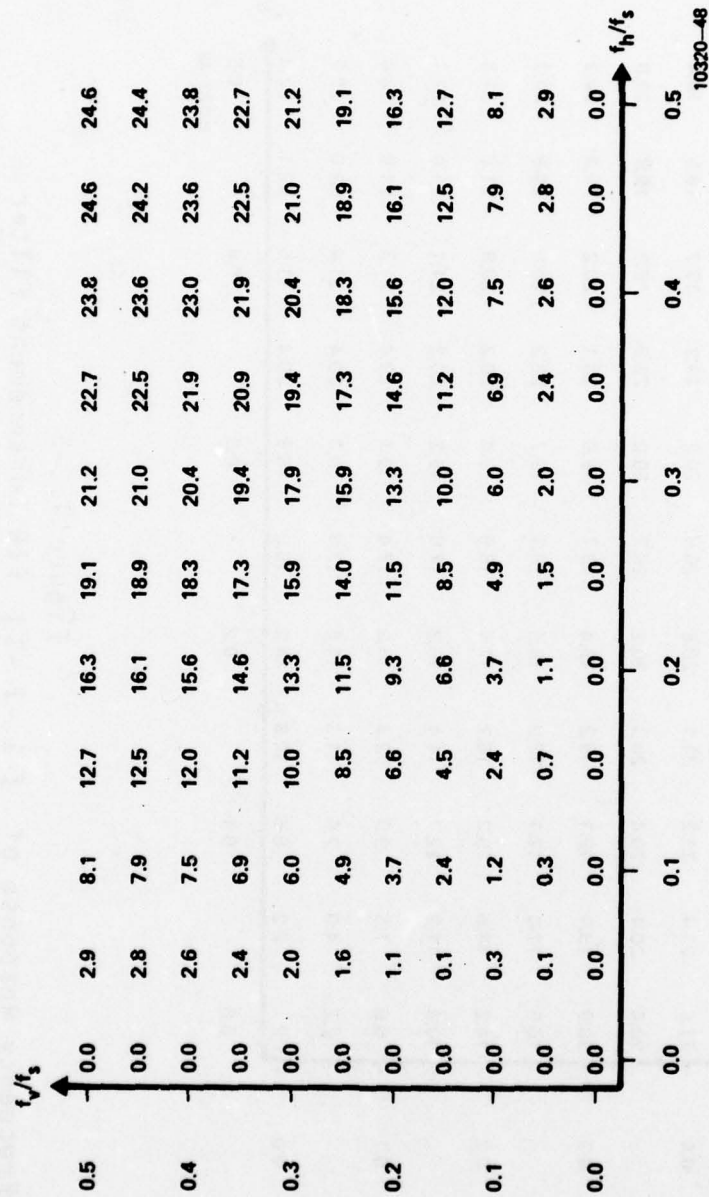


Figure 4.2-4  
Frequency Response of  $\begin{bmatrix} 1 & -2 & 1 \\ -2 & 5 & -2 \\ 1 & -2 & 1 \end{bmatrix}$  FIR Enhancement Filter



The modulation transfer function (MTF) is loosely defined as the magnitude of the frequency response of an imaging system from input to output. MTF correction is defined here as the process of filtering the input range prior to passing through the imaging system so that the effective frequency response is flat. This, in effect, eliminates the image degradation caused by the frequency response of the imagery system. Of course, noise can not be too great; otherwise, its effects will be enhanced by the MTF correction.

In order to give a concrete example of how MTF correction performs, the MTF response of Table 4.3.1 was used as the frequency response to be corrected.

A computer subroutine was incorporated in the computer program described in Section 4.1.5 which reads in the magnitude response, and then generates an FIR filter whose magnitude response approximates the reciprocal of the input. This program allows for the Kaiser parameter to be used to control the stopband attenuation.

Figures 4.3.1 and 4.3.2 are the result of passing the original through a round and square 15x15 FIR filter, generated from the 1D filter of Figure 4.3.8.

As was shown in Section 4.1.3, D/I is accomplished by passing the original image through a LPF: for 2:1 D/I,  $\rho=1/2$  for the ideal LPF needed to prevent aliasing. Suppose that 2:1 D/I is required, and MTF correction is to also be applied. Both of these operations can be applied simultaneously using just the LP decimation filter if this filter has the MTF boost up to its cut-off frequency ( $\rho \approx 1/2$  for 2:1 D/I).

This approach was applied for a 2:1 D/I experiment and boosted frequencies up to the  $\rho=1/2$  point by using



TABLE 4.3.1.  
RESPONSE FOR MTF

<u>Frequency</u>	<u>MTF</u>
0	1.0
1	1.0
2	1.0
3	1.0
4	0.95
5	0.80
6	0.68
7	0.575
8	0.5
9	0.42
10	0.40
11	0.37
12	0.33
13	0.32
14	0.30
15	0.28
16	0.27
17	0.26
18	0.25
19	0.24
20	0.23



Figure 4.3.1  
Round 15x15  $\beta=0.0$   $\rho=1.0$   
MTF Correction

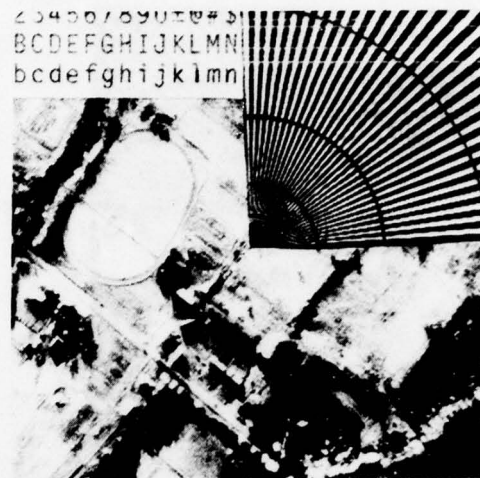


Figure 4.3.2  
Square 15x15  $\beta=0.0$   $\rho=1.0$   
MTF Correction

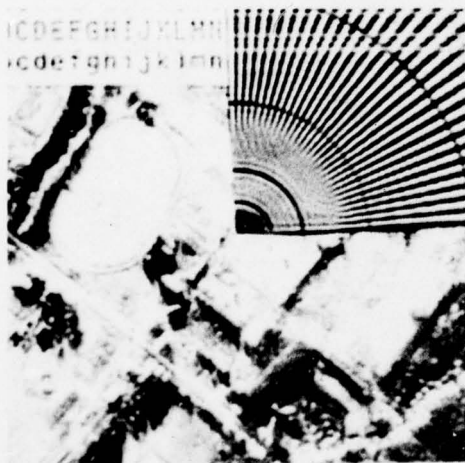


Figure 4.3.3  
2:1 D/I With 6.5 dB  
Frequency Boost.  
15x15 Round Filter  
 $\beta=0.0$   $\rho=0.5$

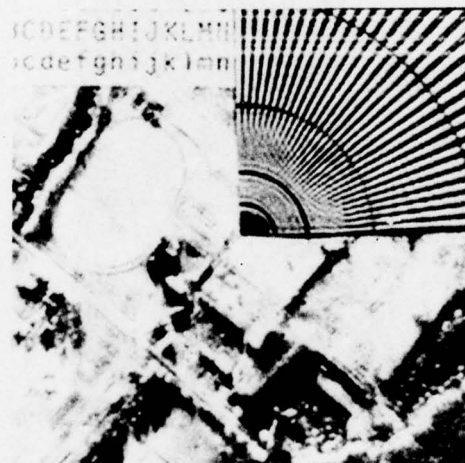


Figure 4.3.4  
2:1 D/I With 6.5 dB  
Frequency Boost.  
15x15 Square Filter  
 $\beta=0.0$   $\rho=0.5$

Figures 4.3.1, 4.3.2, 4.3.3, and 4.3.4

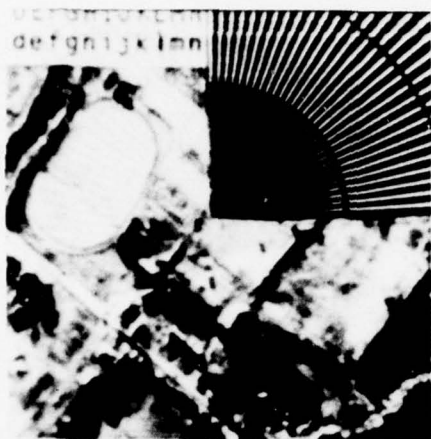


Figure 4.3.5  
2:1 D/I Only 31x31,  
Round  $\beta=1.0$   $\rho=0.46$



Figure 4.3.6  
Applied to Figure 4.3.5

$$\begin{bmatrix} -1 & -1 & -1 \\ -1 & 9 & -1 \\ -1 & -1 & -1 \end{bmatrix}$$



Figure 4.3.7  
Applied to Figure 4.3.5

$$\begin{bmatrix} -2 & -2 & -2 \\ -2 & 17 & -1 \\ -1 & -1 & -1 \end{bmatrix}$$

Figures 4.3.5, 4.3.6 and 4.3.7



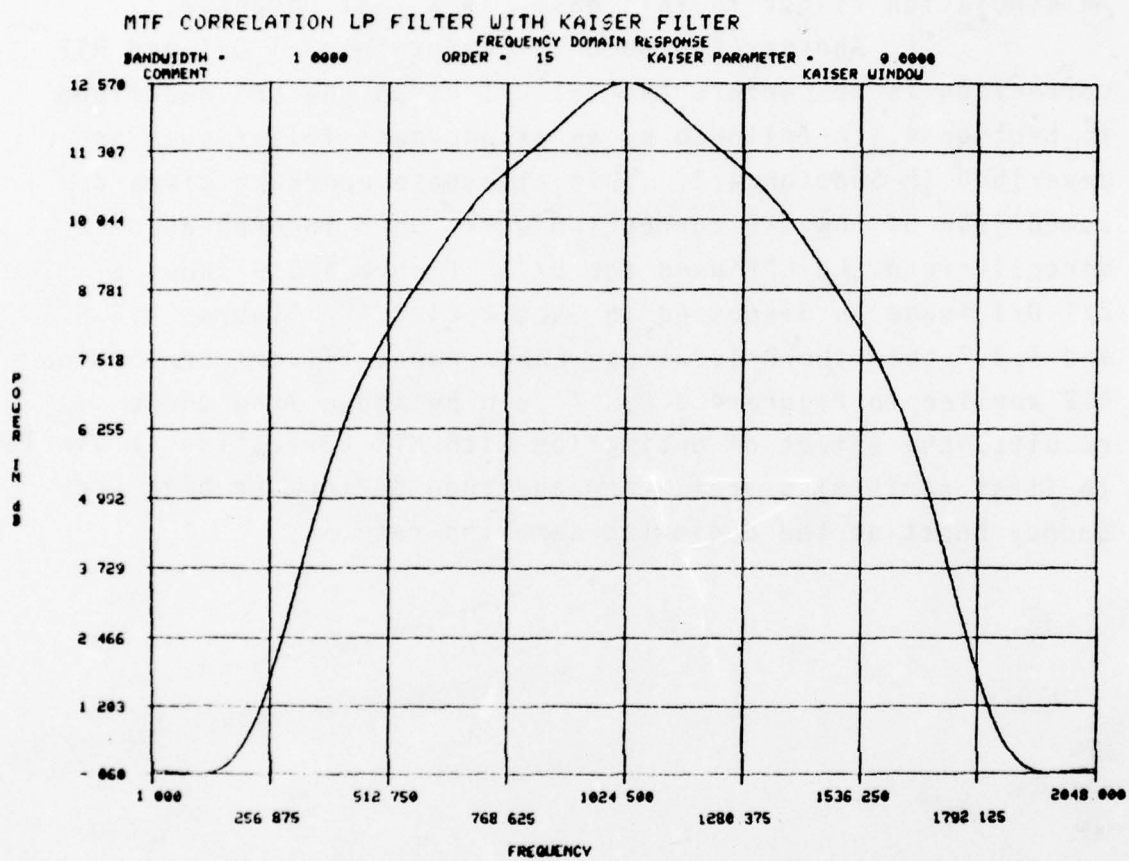


Figure 4.3.8  
 One-Dimensional MTF Filter for Full Resolution

the top half of Table 4.3.1 as the MTF response. The 1D filter generated to perform this MTF correction is shown in Figure 4.3.9 and gives about 6.5 dB boost. This filter was then used to generate a 2D round and square filter, shown in Figures 4.3.10 and 4.3.11. The results of applying these filters are shown in Figures 4.3.3 and 4.3.4. (The interpolation filter in this case has a flat response.)

Another approach to performing 2:1 D/I and MTF correction is to perform the 2:1 D/I using the LPF described in Section 4.1.6 followed by an enhancement filter such as described in Section 4.2. This alternate approach gives a comparison of how MTF correction works when incorporated directly into the LPF used for D/I. Figure 4.3.5 shows a 2:1 D/I image as discussed in Section 4.1.6. Figures 4.3.6 and 4.3.7 show the  $R=1,2$  image enhancement filters of Section 4.2 applied to Figure 4.3.5. As can be shown from these results, the effect of decimation with MTF correction is similar to first performing decimation and then performing high frequency boost at the decimated sampling rate.

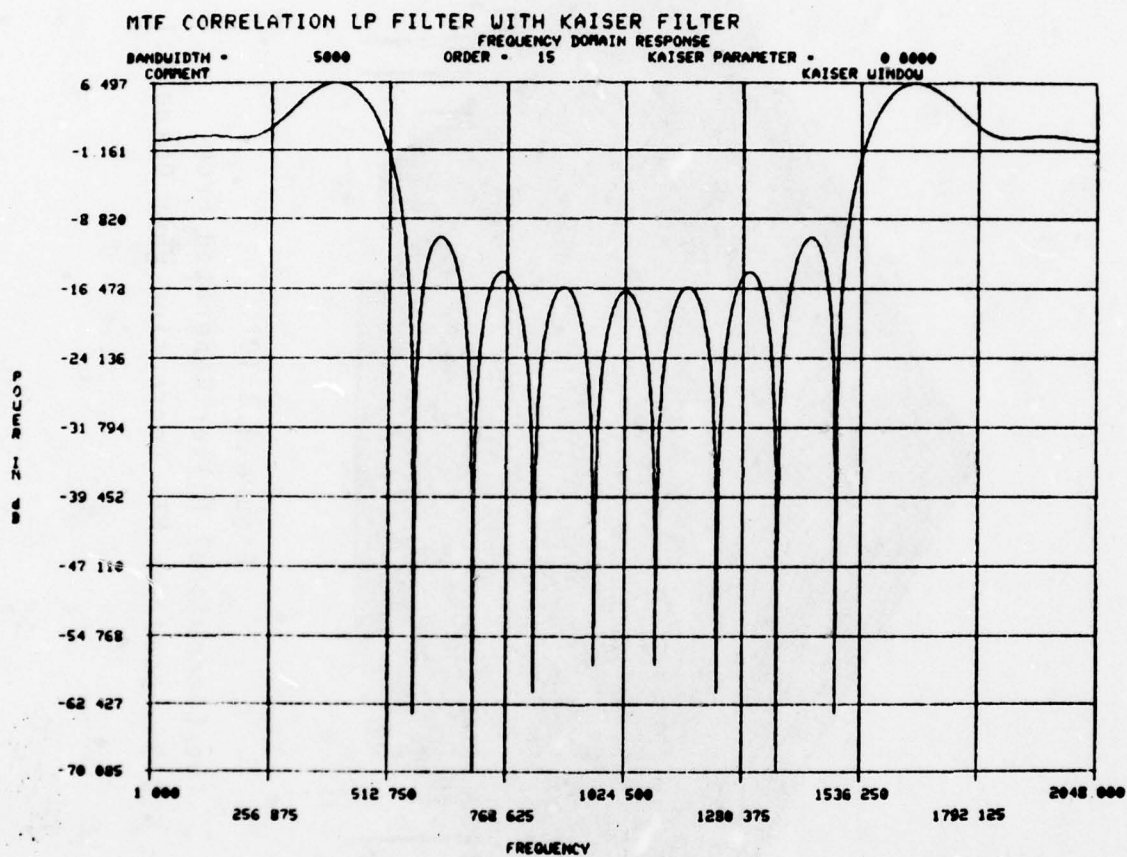


Figure 4.3.9  
 One-Dimensional Filter Used for  
 MTF Correction When Doing 2:1 D/I.



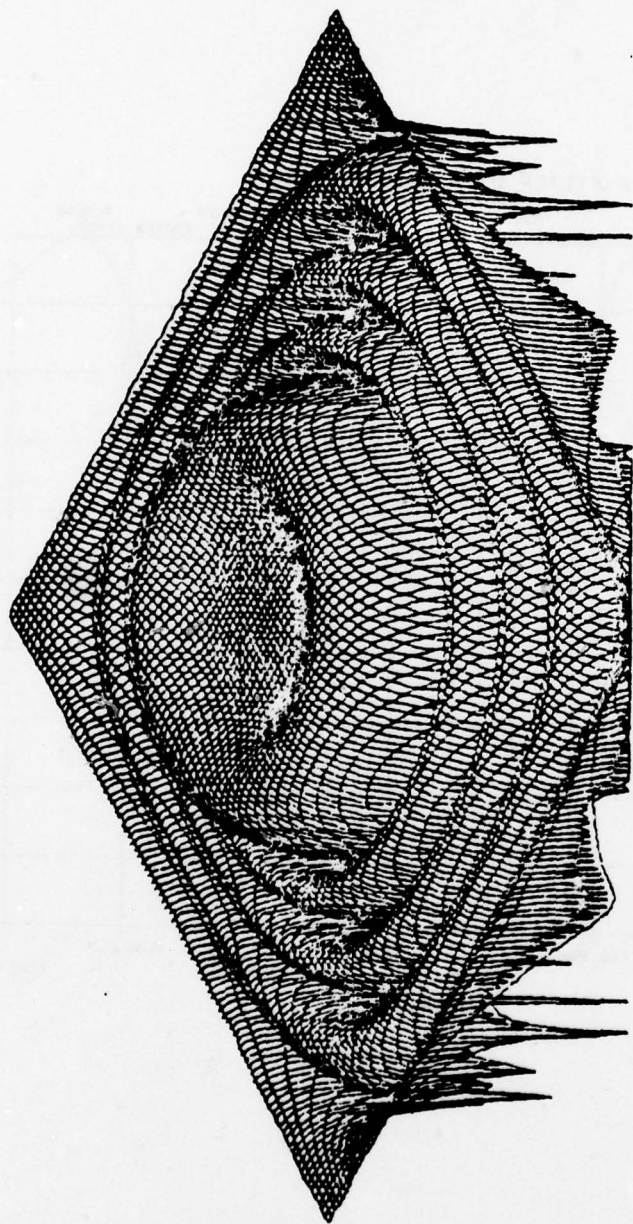


Figure 4.3.10  
Two-Dimensional Filter Generated From  
Figure 4.3.9-- Used for MTF Correction When Doing 2:1 D/I.

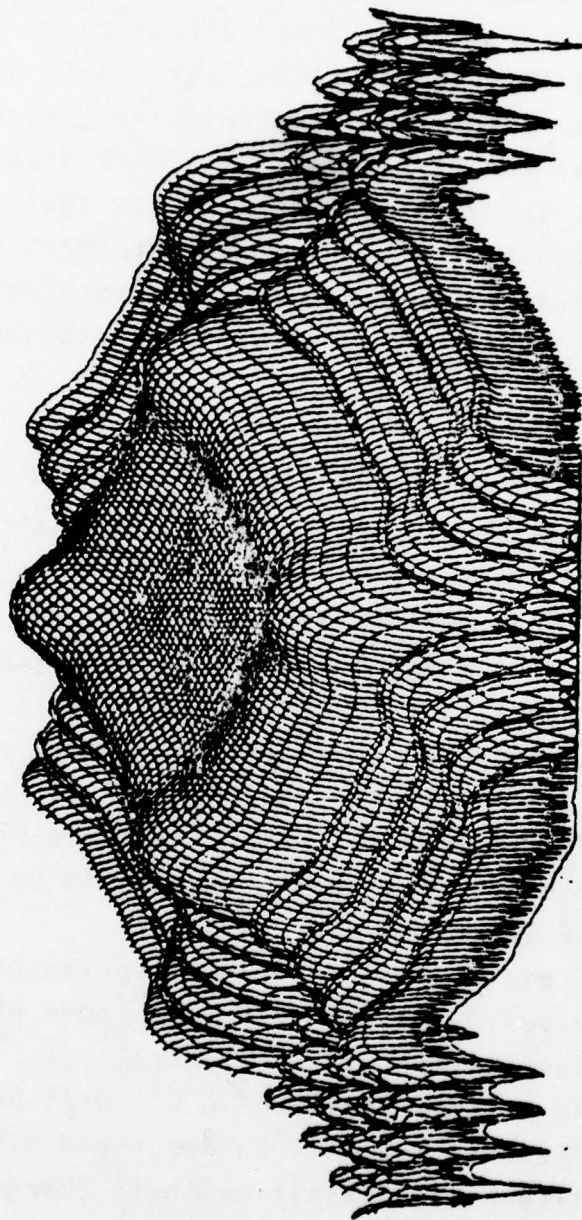


Figure 4.3.11  
Two-Dimensional Filter Generated From  
Figure 4.3.9-- Used for MTF Correction When Doing 2:1 D/I.

#### 4.4

#### PRACTICAL FIR FILTER IMPLEMENTATIONS

In this section we begin by surveying currently available CCD's and their operating characteristics. This is followed by a discussion of the degradation of FIR filter performance due to the nonidealness of available CCD's that must be used in any practical implementation. More specifically, we discuss the effects of the signal dispersion, tap weight errors, and quantization accuracy that are associated with practical CCD's used to implement FIR's. We develop a dispersion model and present computed theoretical results that show the effect of capacitive coupling and CTI among CCD stages.

##### 4.4.1

##### Currently Available Charge-Coupled Devices

Charge coupled device (CCD) technology has enjoyed tremendous progress in the last ten years and CCD's are now available that perform important imaging, storage, and analog signal processing functions. Charge-coupled devices may be thought of as discrete-time, continuous amplitude analog shift registers. The analog sample values are stored as packets of electrical charge, each of which is stored in individual potential wells. Multiple-phase clocking signals cause the electrical charge packets to be shifted serially along the length of the register.

In this section we briefly survey commercially available CCD devices, with emphasis upon those which are suitable for implementation of FIR filters.

The Reticon family of CCD's will be of interest in the ASITS application and will be described first. The Reticon product literature classifies their charge transfer



device products into two classes: those using bucket brigade devices (BBD's) and those using charge coupled devices. The differences between the two lie in the details of device structure. The BBD's allow external access to the individual stages of the analog shift register and therefore facilitate implementation of externally programmable FIR filters. The CCD technology is capable of higher sampling rates and higher density devices than BBD technology.

The Reticon TAD-32A is a 32 stage BBD tapped analog delay line with source follower buffers for each stage. The "A" version reduces the odd-even pattern that was characteristic of the TAD-32. External resistors may be attached to each of the buffer outputs to perform the tap weighting function. This device has a dynamic range of 60 dB and can accommodate sampling rates between 1 kHz and 5 MHz. Two or more TAD-32A's may be cascaded to obtain additional stages of tapped analog delay.

The Reticon R5602 is a mask programmed 64 stage BBD transversal (FIR) filter intended for high volume applications. Sampling rates of 1 MHz and typical stopband rejections of 50 dB can be obtained. The tap weights of the FIR filter are implemented by using a split electrode structure. Both the TAD-32A and R5602 BBD's can be used to implement FIR filters; the former is useful for low-volume experimental applications since it is externally programmable, whereas the latter is suitable for high volume production applications since it is mask programmable and does not require the external tap weight resistors.

In addition to CTD's useful in FIR filtering applications, Reticon also makes a much more sophisticated signal processing device that has potential for bandwidth compression applications. The R5601 CCD performs a 512

point DFT using the Chirp-z algorithm. Each R5601 contains four 512-tap split electrode transversal filters, with tap weight patterns corresponding to two cosine chirps and two sine chirps. Two external multiplier functions are required for use with this device in order to perform the DFT; this is shown in Figure 4.4.1. The R5601 operates at rates as high as 100 kHz and exhibits output SNR's in excess of 70 dB. The commercial availability of the R5601 demonstrates that custom designed 512 tap FIR filters could be fabricated if required for the ASITS. The computer simulation experiments indicate that FIR filters with far fewer stages are adequate.

A CCD of great usefulness in bandwidth compression applications is the Hughes CRC 101. This dual function chip is mounted in a 40 pin DIP and performs a 16-point one-dimensional Hadamard transform (or its inverse) in real time at input sampling frequencies up to 10 MHz, and also contains a 16 input parallel-to-serial multiplexer.

Another CCD of potential interest in a digital storage oriented ASITS is the TMS 3064 65,536-bit CCD memory packaged in a 16-pin DIP. This device operates between 1 MHz and 5 MHz.

The above discussion briefly surveyed commercially available CTD's which are potentially applicable to an ASITS design. The two of importance for the remainder of this section are those which are useful in implementing FIR filters, i.e., the TAD-32A and R5602.

#### 4.4.2      Analyses of Nonideal FIR Implementations

This section deals with three ways in which practical CCD's depart from analog shift registers consisting of ideal delay elements with ideal tap weights:

- (1) dispersion effects due to CTI and capacitive coupling among CCD stages

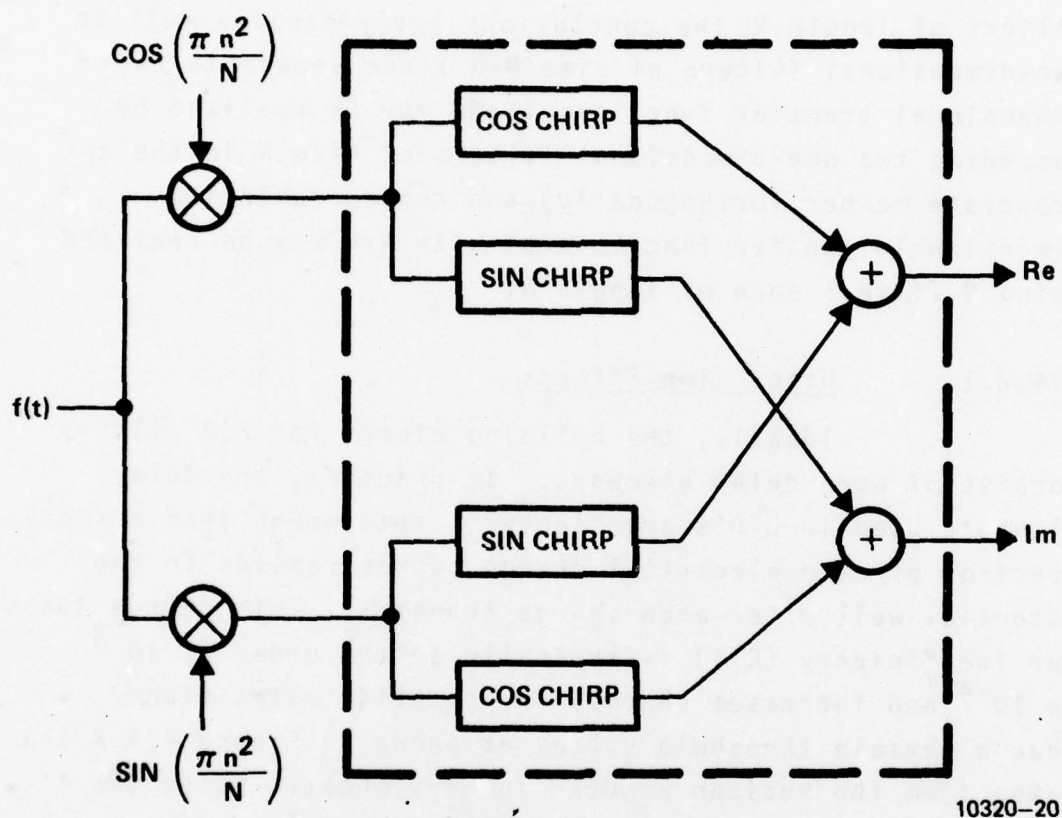


Figure 4.4.1  
Block Diagram of Chirp Z Algorithm



(2) tap weight errors

(3) quantization effects

Although the discussion is with regard to one-dimensional filters of length  $N$  the conclusions apply equally well to two-dimensional filters of size  $N \times N$  since separable two-dimensional transfer functions ( $N \times N$ ) may be realized by cascading two one-dimensional filters of size  $N$  in the appropriate manner (orthogonally) and nonseparable two-dimensional transfer functions of size  $N \times N$  may be realized using  $N$  filters each of length  $N$ .

#### 4.4.2.1 Dispersion Effects

Ideally, the building blocks for FIR filters consist of pure delay elements. In practice, the delay elements used in CTD's are "leaky" - this means that a small fraction of each electrical charge packet remains in the potential well after each charge transfer. This charge transfer inefficiency (CTI) is typically on the order of  $10^{-3}$  to  $10^{-4}$  and increases sharply for sampling rates higher than a certain threshold value, as shown in Figure 4.4.2.1-1 taken from the Reticon product summary sheet. Using the equation describing the charge transfer process [ 9 ]

$$V_k [(n+1)T] = (1-\epsilon) V_{k-1}(nT) + \epsilon V_k(nT) \quad (4-33)$$

we can take  $z$  transforms to obtain

$$V_k(z) = (1-\epsilon)z^{-1} V_{k-1}(z) + \epsilon z^{-1} V_k(z) \quad (4-34)$$

where  $z = \exp(i2\pi f/f_s)$ ,  $f_s$  = sampling rate.

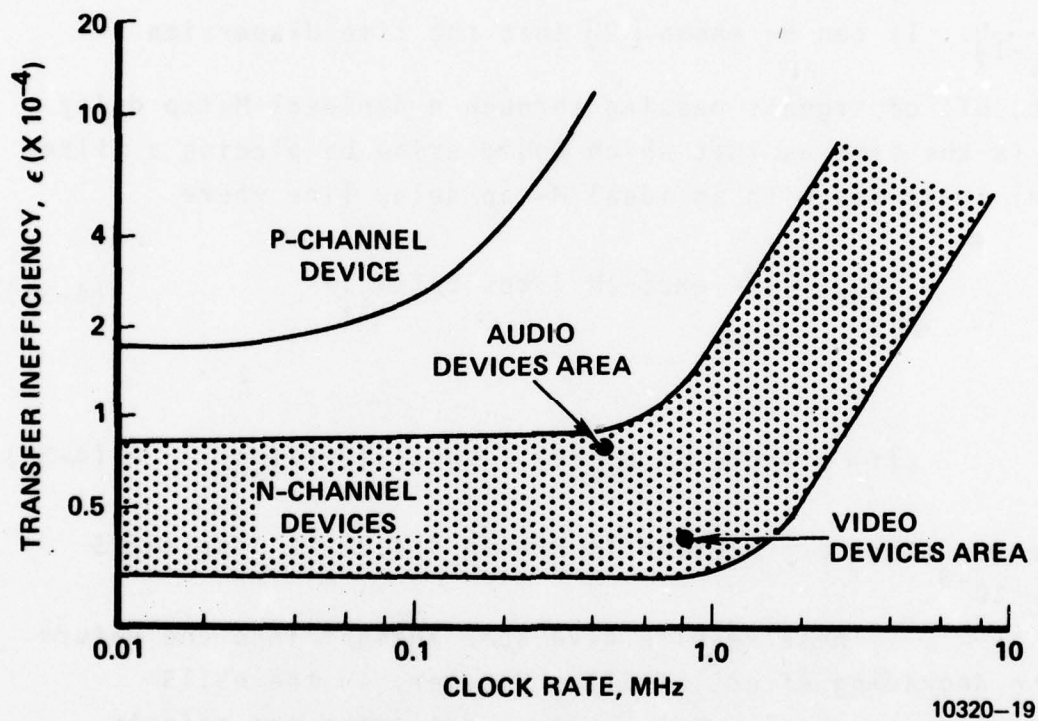


Figure 4.4.2.1-1  
Charge Transfer Inefficiency Versus Clock Frequency

Collecting terms on  $V_k(z)$  in this expression yields

$$V_k(z) = \left( \frac{1-\epsilon}{1-\epsilon z^{-1}} \right) z^{-1} V_{k-1}(z) \quad (4-35)$$

Recalling that  $z^{-1}$  is the  $z$  transform corresponding to a pure delay element, it is clear that the effect of CTI is to modify the  $z$  transform of an ideal delay by the factor  $\left( \frac{1-\epsilon}{1-\epsilon z^{-1}} \right)$ . It can be shown [9] that the time dispersion

due to CTI of signals passing through a nonideal  $M$ -tap delay line is the same as that which would arise by placing a filter  $H_{DL}(z)$  in series with an ideal  $M$ -tap delay line where

$$\left| H_{DL}(f) \right| = \exp\{-\epsilon M [1 - \cos(2\pi f/f_s)]\} \quad (4-36)$$

and

$$\phi(f) = \text{Angle of } H_{DL}(f) = M\epsilon \sin(2\pi f/f_s). \quad (4-37)$$

These functions are plotted in Figure 4.4.2.1-2 for  $M=225$  and  $\epsilon=10^{-3}$ .

These results give some insight into the nature of the degrading effect of CTI. However, in the ASITS application the real question is to determine the effects of CTI upon the magnitude and phase of the lowpass filters used in the decimation and interpolation processes, and the filters used for MTF correction. One way to do this is to substitute the transfer function of the nonideal delay element, denoted by  $(z^{-1})'$ , for the transfer function of the ideal delay element,  $z^{-1}$ , in the expression for  $H(z)$ .



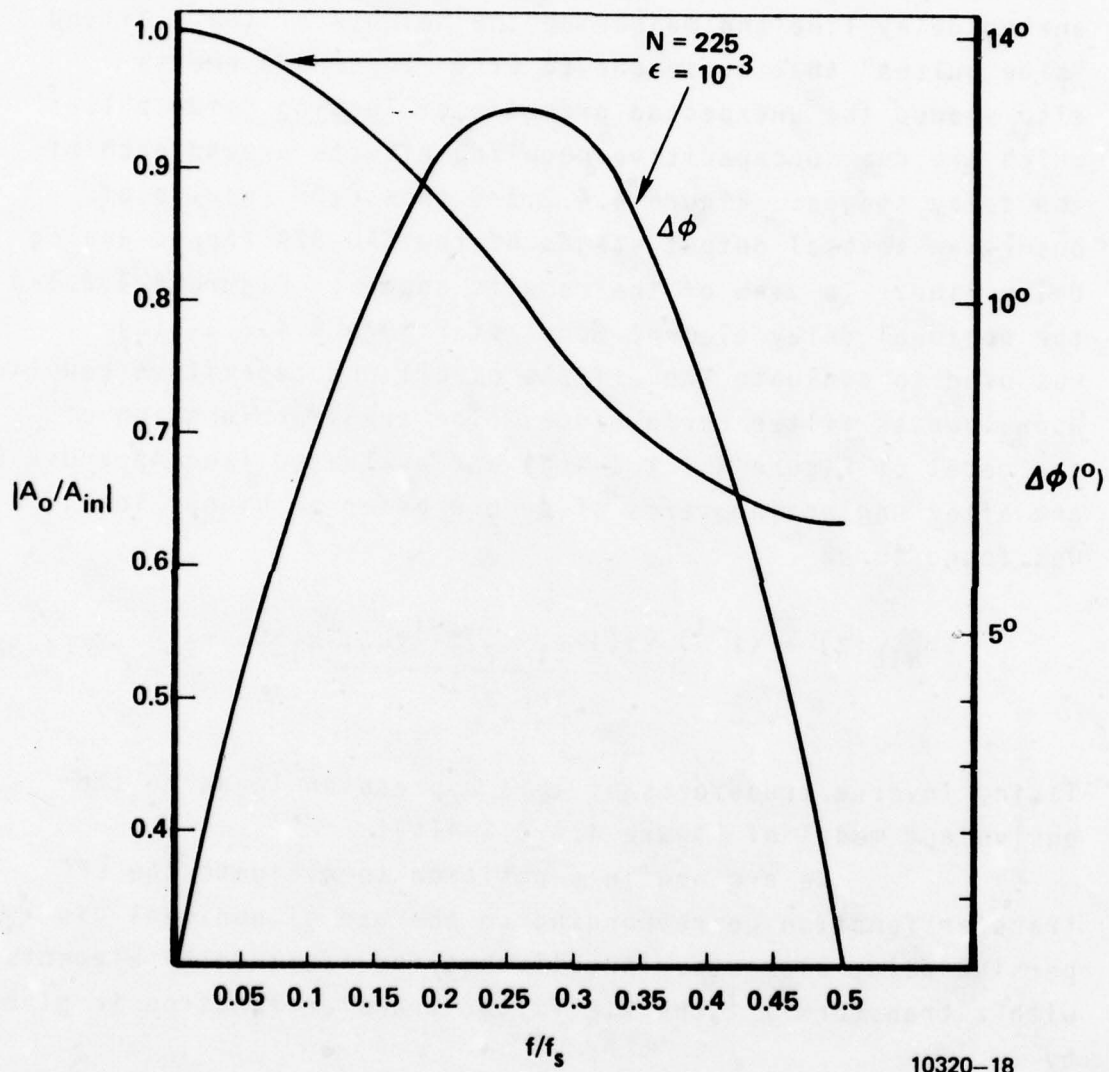


Figure 4.4.2.1-2  
 Effects of CTI on Delay Line Transfer Function

In order to determine the value of the CTI for the Reticon TAD-32A some laboratory measurements were made on this device by injecting a signal into the tapped analog delay line and measuring the heights of the trailing "side pulses" that arose due to CTI. The measurements also showed the unexpected presence of leading "side pulses" which are due to capacitive coupling effects around each of the delay stages. Figure 4.4.2.1-3 shows the results of observing several output stages of the TAD-32A tapped analog delay line. In view of the results shown in Figure 4.4.2.1-3 the nonideal delay element model of Figure 4.4.2.1-4(a) was used to evaluate the effects of CTI and capacitive coupling upon lowpass filter performance. The transfer function of the model of Figure 4.4.2.1-4(a) was evaluated (see Appendix B) and after neglecting terms of second order or higher in  $\epsilon$ , was found to be

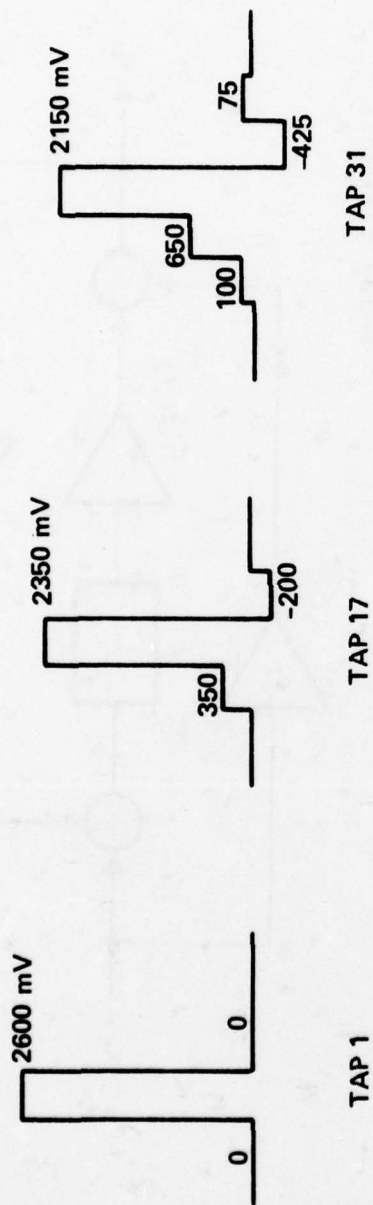
$$H_{NIE}(z) = (z^{-1})' = \frac{(1-\epsilon_1-\epsilon_2)z^{-1}+\epsilon_2}{1-\epsilon_1z^{-1}}, \quad \epsilon_1 \geq 0, \quad \epsilon_2 \leq 0 \quad (4-38)$$

Taking inverse transforms of this expression leads to the equivalent model of Figure 4.4.2.1-4(b).

We are now in a position to evaluate the LPF transfer function corresponding to the use of nonideal dispersive delay elements. Recall that for ideal delay elements with  $z$ -transform  $z^{-1}$  the FIR filter transfer function is given by

$$H_I(z) = \sum_{k=0}^{N-1} h_k z^{-k} \quad (4-39)$$

where the  $h_k$  are the FIR tap weights.



(EVEN TAPS IDENTICAL STRUCTURE, SLIGHTLY HIGHER LEVELS)

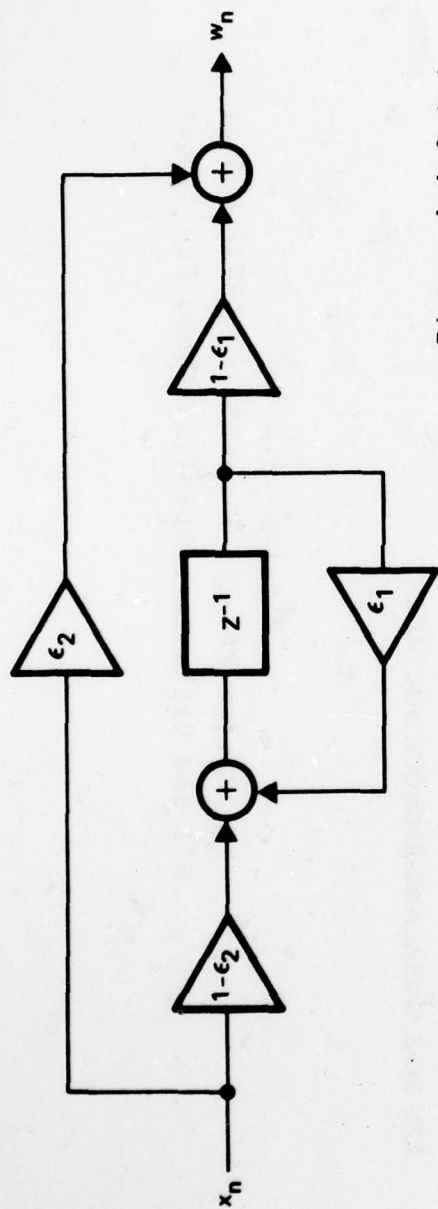
NOTE:

2.3V BIAS, UNBALANCED CLOCK VOLTAGES

10320-17

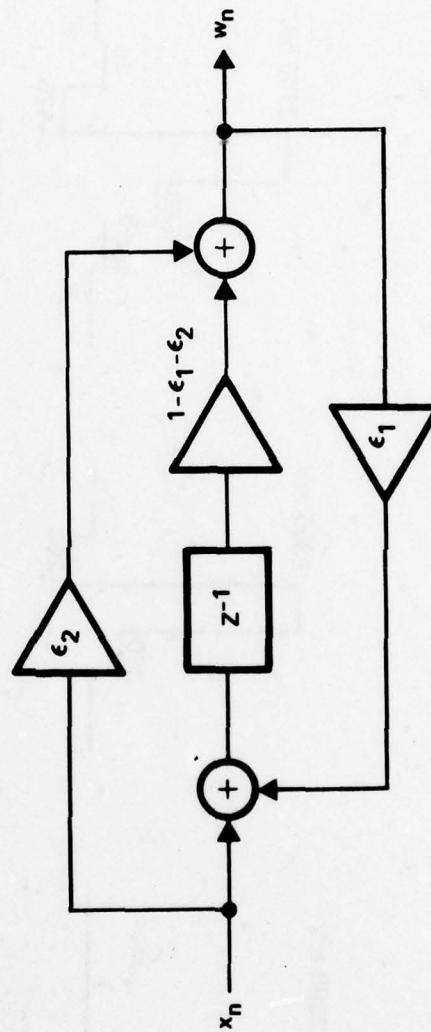
Figure 4.4.2.1-3  
Dispersive Effects Present in TAD-32 Due to  
CTI and Capacitive Coupling Among CCD Stages





a.

Figure 4.4.2.1-4  
Two Equivalent Models for  
Nonideal CCD  
Delay Elements



b.

10320-16

Then for nonideal dispersive delay elements with z-transform

$$(z^{-1})' = \frac{(1-\epsilon_1-\epsilon_2)z^{-1}+\epsilon_2}{1-\epsilon_1 z^{-1}} \quad (4-40)$$

the transfer function is given by

$$H_{NI}(z) = \sum_{k=0}^{N-1} h_k \left[ \frac{(1-\epsilon_1-\epsilon_2)z^{-1}+\epsilon_2}{1-\epsilon_1 z^{-1}} \right]^k \quad (4-41)$$

Since there appear to be no results in the literature regarding simultaneously present "forward" and "backward" dispersion in delay lines used to implement FIR filters, this expression was evaluated using a computer with a graphics display capability. The tap weights  $h_k$  were computed as the product of a sinc function and a Hamming window function:

$$h_k = \left[ 0.54 + 0.46 \cos\left(\frac{2\pi}{N}\left(k+1-\frac{N+1}{2}\right)\right) \right] \times \quad (4-42)$$

$$\left[ \sin\left(\pi \left(k+1-\frac{N+1}{2}\right)\right) / \left(\pi \left(k+1-\frac{N+1}{2}\right)\right) \right] (fc/\pi) \quad (4-43)$$

for  $k=0,1,2,\dots,N-1$  where  $N$  is an odd number

The sinc function impulse response using a unity window gives only 13 dB of stopband attenuation; using a Hamming window on the sinc function gives about 40 dB additional stopband attenuation at the expense of increased filter transition bandwidth. The resulting stopband attenuation of approximately 53 dB is evident in all of the filter transfer functions that were obtained.

This computer program was exercised using a variety of values for  $\epsilon_1$ ,  $\epsilon_2$ , and  $N$  for two values of  $f_c$ , the filter cutoff frequency. The graphics display for each run includes plots of the filter frequency transfer function using both linear and logarithmic (dB) ordinate scales, the filter phase response, and the filter impulse response.

Since the degradation of the linear phase response in these plots was imperceptible, the phase response plots are not shown in the figures below. The impulse response plots were used to verify the filter design portion of the program and consequently are not shown in all of the figures below. Some representative results are shown in Figures 4.4.2.1-5 through 4.4.2.1-9 for the  $N=31$  case and additional results are contained in Appendix C for  $N=9$ ,  $N=15$ ,  $N=63$ , and  $N=225$ . Choosing 0.01 as the maximum magnitude of  $\epsilon_1$  and  $\epsilon_2$  represents the worst case situation likely to be encountered. Typical values for  $\epsilon_1$  (CTI) range from  $10^{-4}$  to  $10^{-3}$  according to the literature; Figure 4.4.2.1-1 shows that a well-defined knee exists in the curve relating  $\epsilon_1$  to clock frequency. Experimental measurements using a Reticon TAD-32 at a clock rate of approximately 1 MHz yielded a value a value of  $\epsilon_1=0.0092$  and  $|\epsilon_2| = 0.0057$ . The details regarding the derivation of these results are contained in Appendix B.

Turning now to the computed theoretical results based upon the model of Figure 4.4.2.1-4, we first display the ideal case ( $\epsilon_1=\epsilon_2=0$ ) in Figure 4.4.2.1-5 for comparison with subsequent results. The "(a)" portion of this and subsequent figures utilizes a linear ordinate, whereas the corresponding "(b)" portions utilize a logarithmic ordinate (dB scale). The abscissa in each case is frequency normalized by the sampling rate  $f_s$ . Note that the filter cutoff frequency  $f_c$  is normalized by  $f_s/2$ . Figure 4.4.2.1-6 shows



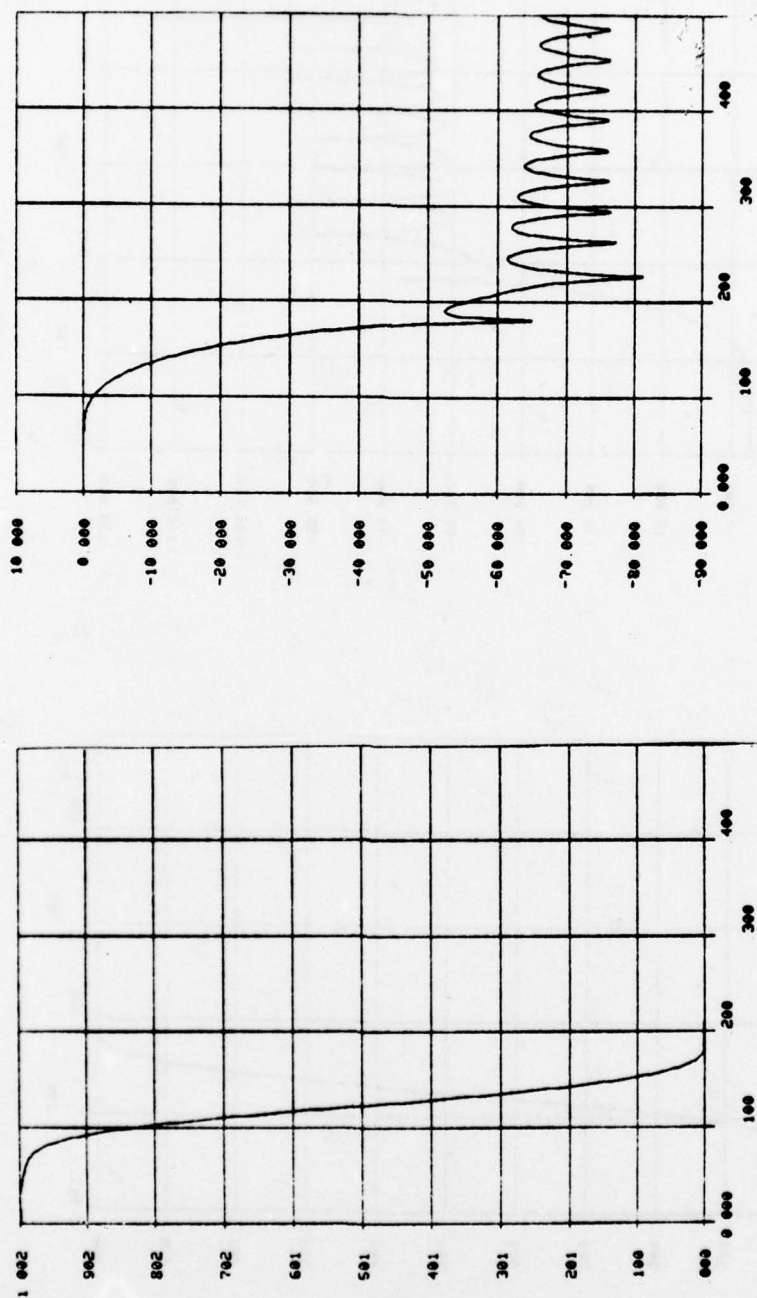


Figure 4.4.2.1-5  
 $N=31$ ,  $\epsilon_1=0$ ,  $\epsilon_2=0$ ,  $f_c=.25(f_s/2)$

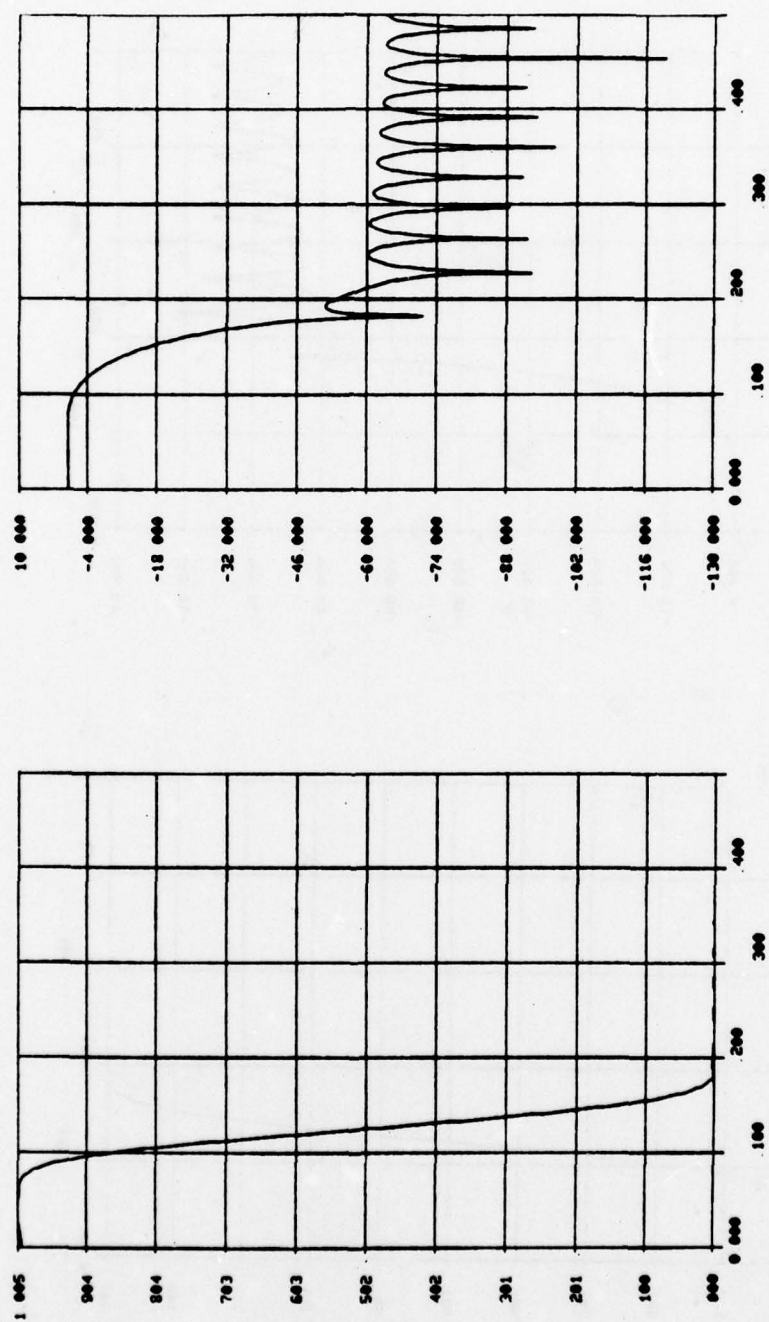


Figure 4.4.2.1-6  
 $N=31$ ,  $\epsilon_1=.01$ ,  $\epsilon_2=0$ ,  $f_c=0.25(f_s/2)$

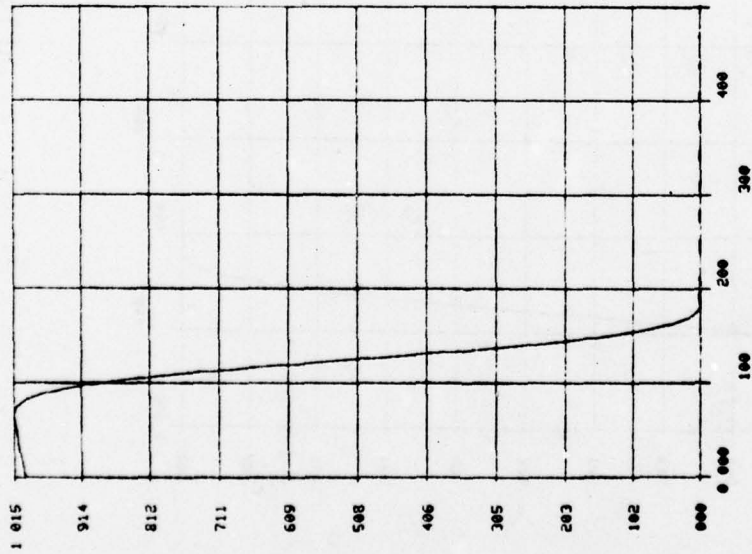
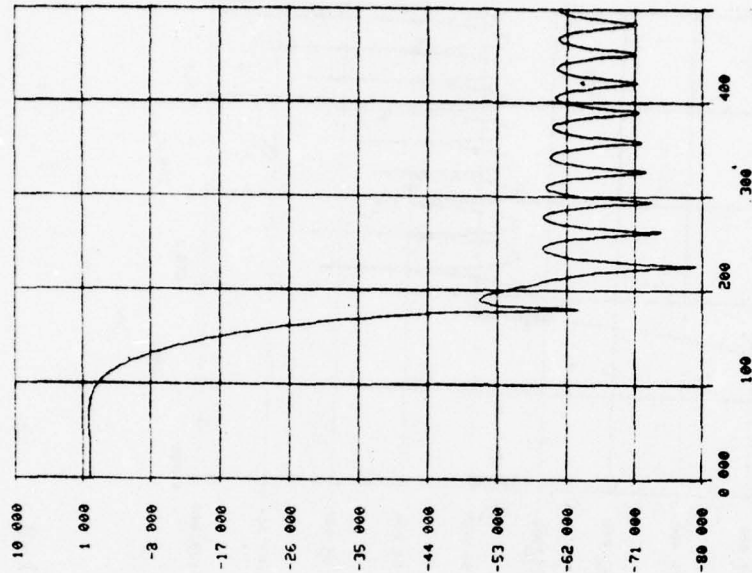


Figure 4.4.2.1-7  
 $N=31$ ,  $\epsilon_1=0$ ,  $\epsilon_2=-.01$ ,  $f_c=0.25$  ( $f_s/2$ )



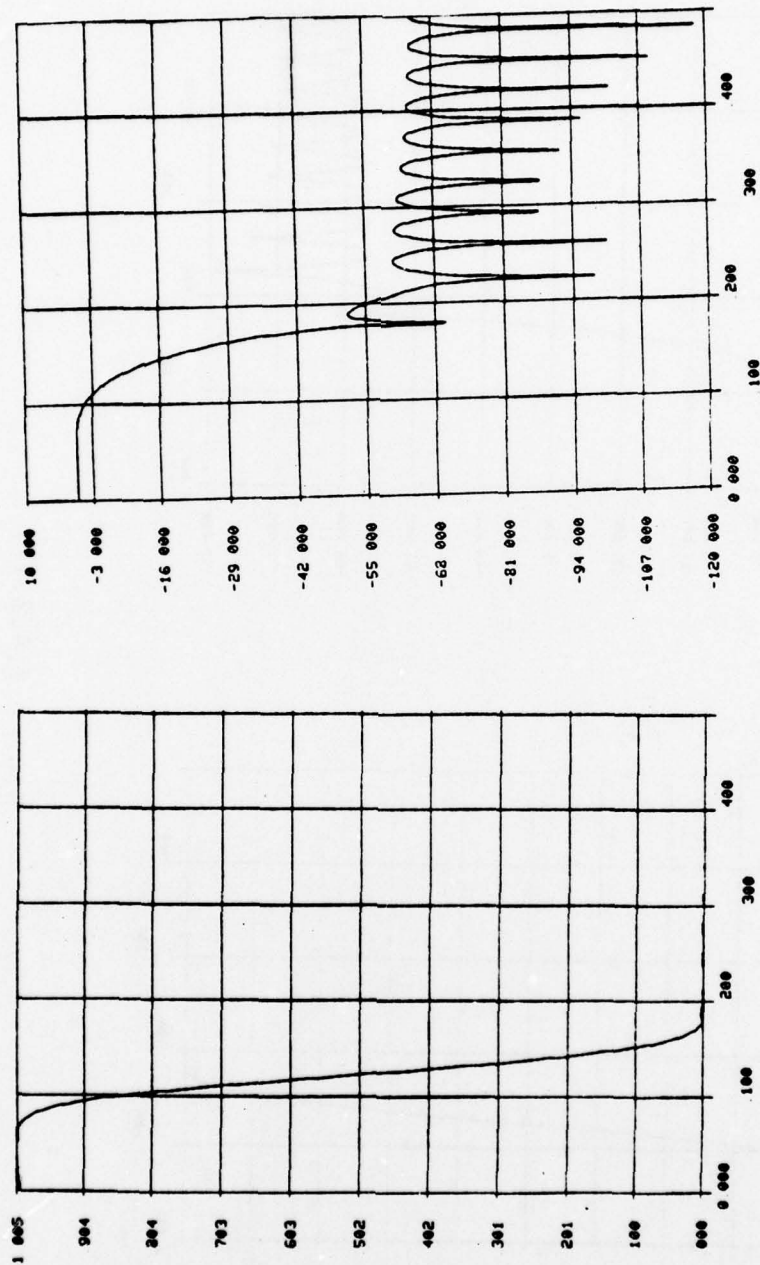


Figure 4.4.2.1-8  
 $N=31$ ,  $\varepsilon_1=.01$ ,  $\varepsilon_2=-.01$ ,  $f_c=0.25(f_s/2)$

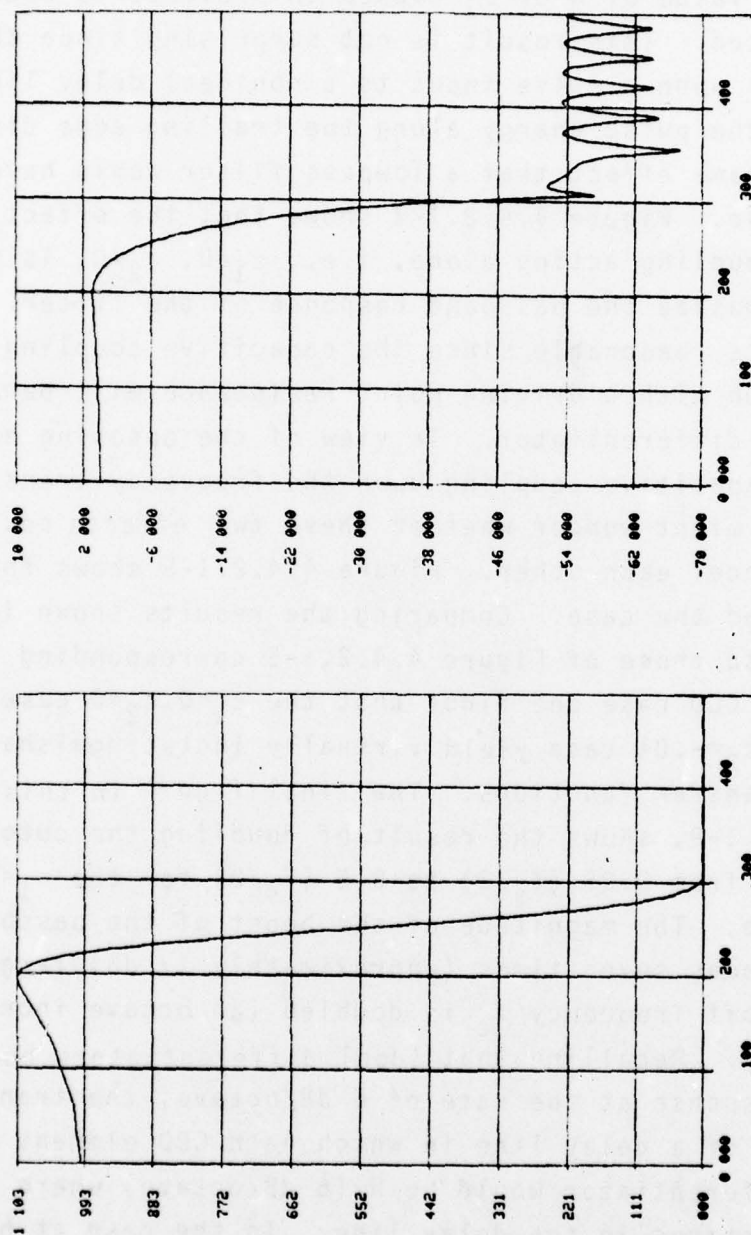


Figure 4.4.2.1-9  
 $N=31$ ,  $\epsilon_1=0$ ,  $\epsilon_2=-.01$ ,  $f_c=0.5$  ( $f_s/2$ )

that the effect of CTI acting above, i.e.,  $\epsilon_1 \neq 0$ ,  $\epsilon_2 = 0$ , is to cause the passband response of the filter to droop; increasing the value of  $N$  or  $\epsilon_1$  causes this effect to become more pronounced. This result is not surprising since the effect of CTI upon a pulse input to a nonideal delay line is to smear the pulse energy along the trailing edge direction-- this is the same effect that a lowpass filter would have upon an input pulse. Figure 4.4.2.1-7 shows that the effect of capacitive coupling acting alone, i.e.,  $\epsilon_1 = 0$ ,  $\epsilon_2 \neq 0$ , is to boost or emphasize the passband response of the filter. Again, this is reasonable since the capacitive coupling in conjunction with a driving point resistance will behave like a leaky differentiator. In view of the opposing effects of CTI and capacitive coupling upon the frequency transfer function one might wonder whether these two effects could partially cancel each other. Figure 4.4.2.1-8 shows that this is indeed the case. Comparing the results shown in this figure to those of Figure 4.4.2.1-5 corresponding to the ideal CCD case one finds that the  $\epsilon_1 = 0, \epsilon_2 = 0$  case and the  $\epsilon_1 = .01, \epsilon_2 = -.01$  case yield virtually indistinguishable frequency transfer functions. The final figure in this series, Figure 4.4.2.1-9, shows the result of doubling the cutoff frequency  $f_c$  from  $0.25 (f_s/2)$  to  $0.5 (f_s/2)$  for the  $\epsilon_1 = 0, \epsilon_2 = -0.01$  case. The magnitude of the boost of the passband response becomes seven times (approximately 17 dB) larger when the cutoff frequency  $f_c$  is doubled (an octave increase in frequency). Recalling that ideal differentiators boost frequency response at the rate of 6 dB/octave, the transfer function of a delay line in which each CCD element was an ideal differentiator would be  $N \times (6 \text{ dB/octave})$  where  $N$ =number of stages in the delay line. In the case at hand the differentiation is "leaky" and the frequency transfer function is that of a generalized filter rather than a delay line.



All of the observations that have been made regarding the  $N=31$  FIR filters also apply to filters with a greater or smaller number of stages. The effects discussed above are more pronounced for larger values of  $N$  and less pronounced for smaller values of  $N$ , as may be seen by referring to the additional results in Appendix C.

As a result of these computations based upon a model which includes the effects of either CTI or capacitive coupling, or both simultaneously, we make the following observations regarding FIR filters implemented with practical CCD's exhibiting time dispersion.

- (1) the filter passband response:
  - droops if  $\epsilon_1 \neq 0, \epsilon_2 = 0$
  - peaks if  $\epsilon_1 = 0, \epsilon_2 \neq 0$
  - is largely unaffected if  $\epsilon_1 \neq 0, \epsilon_2 \neq 0, |\epsilon_1| = |\epsilon_2|$
- (2) the phase response is not appreciably affected regardless of the combination of  $\epsilon_1$  and  $\epsilon_2$  values for  $\epsilon_1 < 0.01, |\epsilon_2| < 0.01$
- (3) Minimum stopband attenuation is unaffected
- (4) nulls in the stopband response are partially filled in for  $\epsilon_1 = 0, \epsilon_2 \neq 0$  or  $\epsilon_1 \neq 0, \epsilon_2 = 0$ . This is of no practical consequence.
- (5) transition bandwidth is unaffected
- (6) the effects of CTI and capacitive coupling tend to offset each other

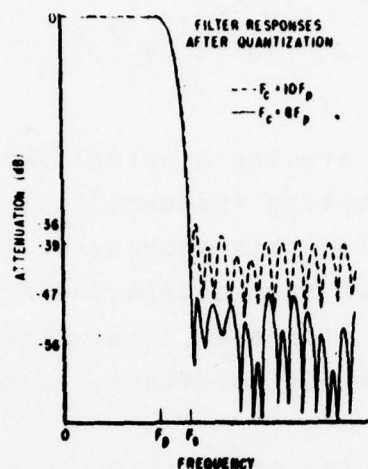
On the basis of these worst case simulation results we conclude that commercially available CCD's are adequate to meet the ASITS filtering requirements. For FIR filters with a large number of taps the filter weights can be modified to counteract the droop in the passband response by designing a "pre-emphasis" filter assuming ideal CCD elements with just enough passband boost to compensate for the passband droop due to CTI dispersion along the delay line.

#### 4.4.2.2 Tap Weight Tolerance and Quantization Effects

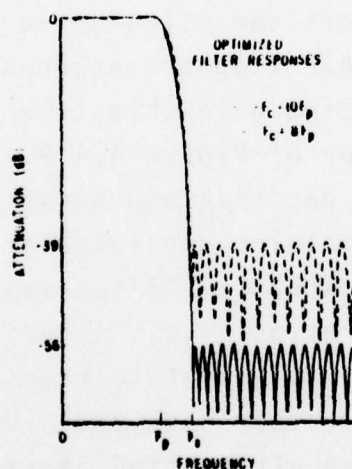
Two issues of some importance in practical realizations of FIR filters are random tap weight errors and tap weight quantization errors. These issues and others have been studied by Baertsch et al [10] and their results are applicable to the ASITS application. Random tap weight errors are due to CCD process variations such as nonuniform etching whereas quantization errors arise as a result of computerized photolithographic maskmaking processes in which the tap weights must be rounded off to integral multiples of a minimum coordinate spacing.

The effect of quantization of tap weights to one part in 600 is shown in Figure 4.4.2.2-1 (taken from [10]) for an optimum equiripple FIR filter designed using the Parks-McClellan algorithm [11].

It is clear that the passband response and transition bandwidth are not appreciably affected, but that the stopband attenuation is substantially degraded by 3 dB to 9 dB depending upon relative sampling frequency. Figure 4.4.2.2-1 also demonstrates the value of choosing the sampling rate to be as low as possible when designing FIR filters, since this decreases the required filter order by virtue of the empirical relationship due to Rabiner et. al. [13]



(a) Calculated Filter Response After Quantization to One Part in 600



(b) Calculated Filter Response for Two Optimized Filters. Passband and Transition Band Specifications are Identical, But Ratio of Clock Frequencies is 5/4

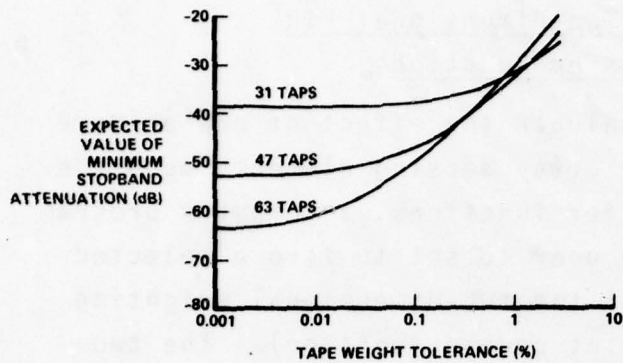


$$N = \frac{-10 \log_{10}(\delta_s \delta_p) - 15}{14 (f_s - f_p)/f_{cl}} \quad (4-44)$$

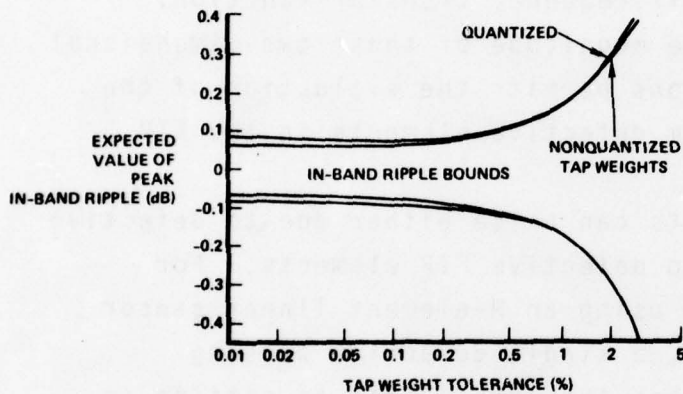
where  $f_s$ ,  $f_p$ , and  $f_{cl}$  are the unnormalized stopband, passband, and clock or sampling frequencies, respectively.

By solving this expression for  $\delta_2$  (stopband ripple) it can be shown that decreasing  $f_{cl}$  allows  $\delta_2$  to be decreased (increasing stopband attenuation) while maintaining all other filter parameters constant, as shown in Figure 4.4.2.2-1.

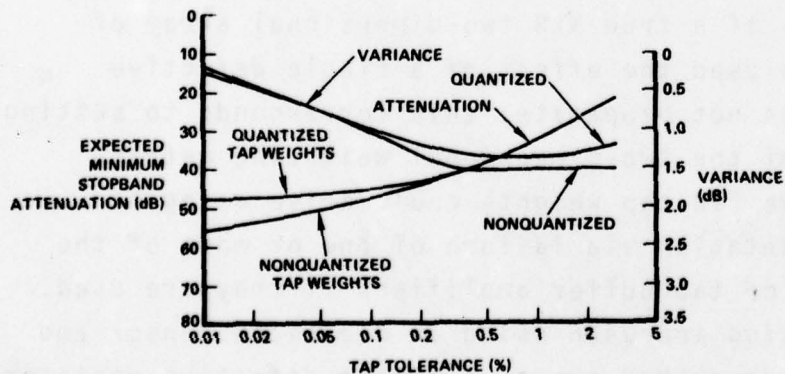
The results of Monte Carlo tap weight tolerance calculations performed by Baertsch *et. al.* [10] are shown in Figure 4.4.2.2-2. The (a) portion of this figure shows that increasing the number of taps in a FIR filter decreases the permissible tap weight tolerance if degradation of stopband attenuation is to be avoided; the  $N=31$  FIR filters discussed elsewhere in this report can allow up to 1% tap weight tolerance without appreciable stopband attenuation degradation, whereas the corresponding allowable tolerance for  $N=63$  is only 0.01%. The (b) portion of Figure 4.4.2.2-2 shows how the average value of peak positive and negative passband ripples varies with tap weight tolerance (with and without tap weight quantization) for the  $N=63$  FIR filter case; if passband ripple were the only consideration (i.e., ignoring stopband attenuation degradation) the tap weight tolerance could be chosen as high as 0.5 to 1%, rather than the 0.01% value required for negligible stopband attenuation degradation. The (c) portion of Figure 4.4.2.2-2 displays the mean and variance of stopband attenuation versus tap weight tolerance for both the quantized and unquantized tap weight case. This result shows that a reasonable choice for tap weight tolerance is in the 0.1% to 0.3% range.



a. MONTE CARLO CALCULATION OF EXPECTED VALUE OF MINIMUM STOP-BAND ATTENUATION VERSUS TAP WEIGHT TOLERANCE FOR THREE OPTIMIZED LOW-PASS FILTER DESIGNS



b. MONTE CARLO CALCULATION OF EXPECTED VALUE OF PEAK IN-BAND RIPPLE VERSUS TAP WEIGHT TOLERANCE FOR 63-TAP LOW-PASS FILTER OF a. IN ONE CASE, TAP WEIGHTS WERE QUANTIZED TO 1 PART IN 600



c. MONTE CARLO CALCULATION OF EXPECTED VALUE AND VARIANCE OF MINIMUM STOPBAND ATTENUATION VERSUS TAP WEIGHT TOLERANCE FOR 63-TAP LOW-PASS FILTER OF a. QUANTIZATION WAS TO 1 PART IN 600

Figure 4.2.2.2-2  
The Effect of FIR Tap Weight  
Error Tolerance Upon  
Filter Response

#### 4.4.3

#### The Effect of Defective FIR Filter Elements Upon Two-Dimensional FIR Frequency Transfer Functions

In order to evaluate the effect of one or more defective FIR tap weights or array sensing elements upon the two-dimensional filter transfer functions, a computer program was written which allows the user to set to zero a selected element or row of elements of the two-dimensional weighting matrix (i.e., the sampled point spread function). The two-dimensional DFT of the resulting defective weighting matrix yields the two-dimensional frequency transfer function. Perspective display of the magnitude of these two-dimensional frequency transfer functions permits the evaluation of the degradation resulting from defective elements in the FIR weighting matrix.

Array defects can arise either due to defective sensing elements or due to defective FIR elements. For an array sensing approach using an N-element linear sensor array and an  $N^2$ -stage TAD, a single defective sensing element with zero output has the same effect as setting an entire row of the two-dimensional weighting matrix to zero. On the other hand, if a true NxN two-dimensional array of sensor elements is used the effect of a single defective sensor element does not propagate; this corresponds to setting a single element of the two-dimensional weighting matrix to zero. Defective FIR tap weights could arise using a TAD-32 FIR implementation via failure of one or more of the tap resistors and/or tap buffer amplifiers if they are used. For the array sensing approach using an N-element sensor and an  $N^2$  element TAD described above, a single defective resistor now is equivalent to setting a single element of the  $N^2$



weighting matrix to zero; recall that a single defective sensing element in this approach is equivalent to setting an entire row of the weighting matrix to zero. For separable FIR filter realizations utilizing a cascade of two one-dimensional FIR filters a single defective resistor is equivalent to setting an entire row or column of the weighting matrix to zero. Separable and nonseparable two-dimensional FIR filters were discussed in Section 4.1.5.

Since setting an element of the weighting matrix to zero is equivalent to adding an impulse of opposite sign and strength at the appropriate location of the two-dimensional weighting matrix, we can infer the following effects of a defective element upon the frequency transfer function:

- (1) change in gain at the origin of the frequency transfer function, i.e.,  $|H(0,0)| \neq 0$  dB
- (2) nonlinear phase of the filter (unless the defective element is the center one of the PSF, odd N)
- (3) distortion of the magnitude response of the transfer function
- (4) decreased stopband attenuation of the transfer function

Effect (1) arises since

$$H(0,0) = \sum_{k=1}^{N-1} \sum_{j=1}^{N-1} h_{jk=1}$$

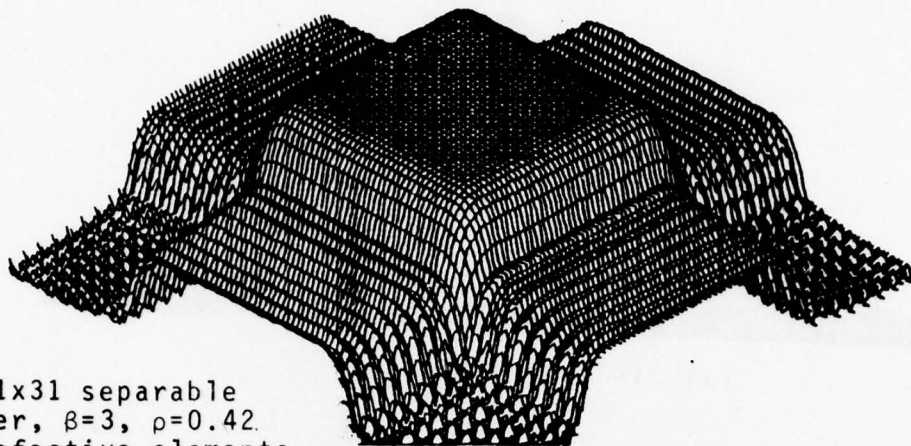
(where  $h_{jk}$  are the filter tap weights) and setting one of the  $h_{jk}$  to zero will prevent the double summation over  $h_{jk}$  from attaining its desired value of unity. Effect (2) is due

to the fact that linear phase FIR filters require tap weight symmetry with respect to the midpoint of the sampled point spread function. This symmetry is destroyed if any tap weight element other than the center one (for odd  $N$ ) is changed. Effect (3) is an obvious result of modifying an element of the tap weight matrix; the extent of the distortion depends upon the magnitude and location of the defective element. The decreased stopband attenuation, effect (4), may be regarded as the result of adding an impulse to the tap weight matrix; superposition applies and the DFT of an impulse is two-dimensional white noise. If the magnitude of this "white noise" exceeds the stopband attenuation, filter performance will be degraded.

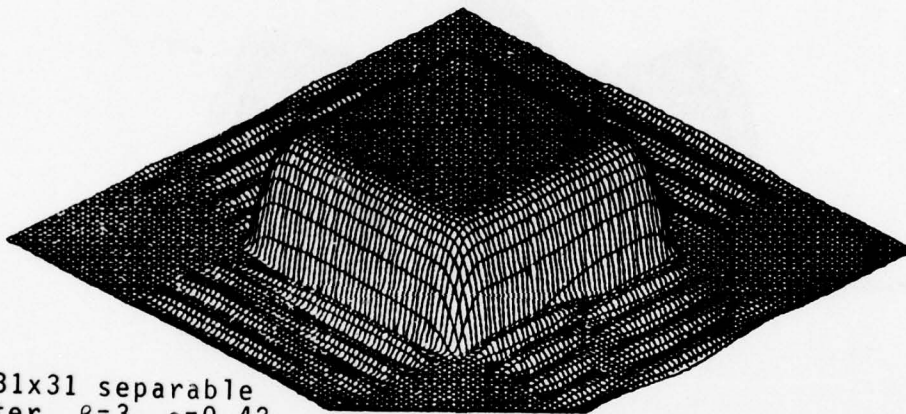
Note that the term "white noise" is somewhat misleading as used in this explanation; it is intended only to help describe the shape of the transfer function in the stopband region and is not to be interpreted as being additive noise in any sense.

Figures 4.4.3-1 through 4.4.3-6 show the perspective plots of the magnitude of the two-dimensional transfer functions that result when various elements or rows of elements of the filter tap weight matrix are set equal to zero.

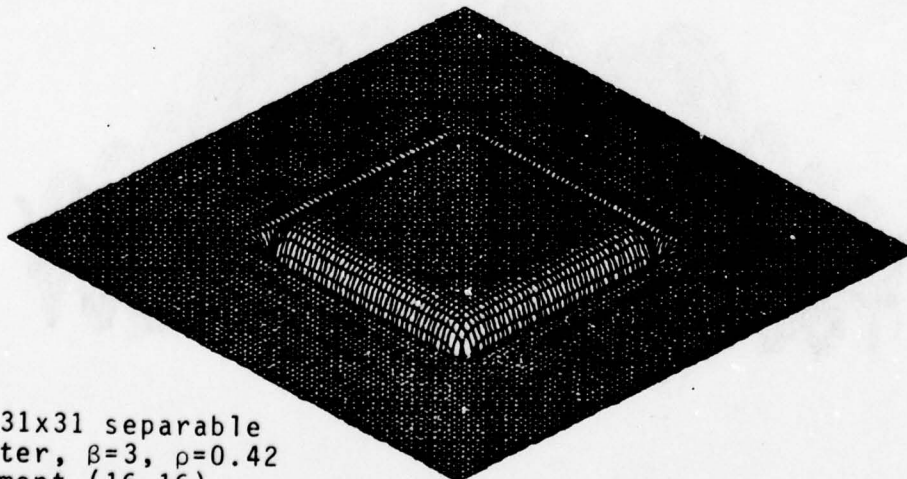
Figure 4.4.3-1(a) shows the magnitude of the transfer function for a  $31 \times 31$  filter without any defective elements; this separable filter is characterized by Kaiser parameter  $\beta=3$  and bandwidth parameter  $\rho=0.42$ . The (b) portion of the figure shows that setting element (15, 18) to zero does not hamper the effectiveness of the filter, whereas Figure 4.4.3-1(c) shows that losing the center filter element (16, 16) renders this half bandwidth filter useless. Recall that deleting (i.e., setting to zero) an element of the filter matrix lowers the gain and raises the stopband; both of these effects are evident in Figure 4.4.3-1.



(a) 31x31 separable  
filter,  $\beta=3$ ,  $\rho=0.42$   
no defective elements



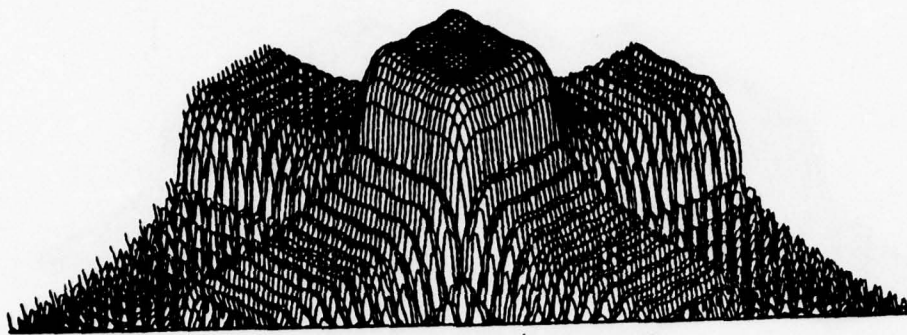
(b) 31x31 separable  
filter,  $\beta=3$ ,  $\rho=0.42$   
element (15,18)  
defective



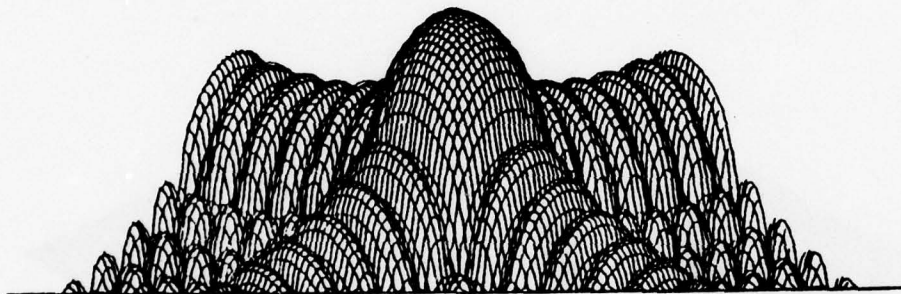
(c) 31x31 separable  
filter,  $\beta=3$ ,  $\rho=0.42$   
element (16,16)  
defective

Figure 4.4.3-1. The Effect  
of a Defective Element  
Upon a Separable 31x31,  
 $\beta=3.0$ ,  $\rho=0.42$  FIR Filter

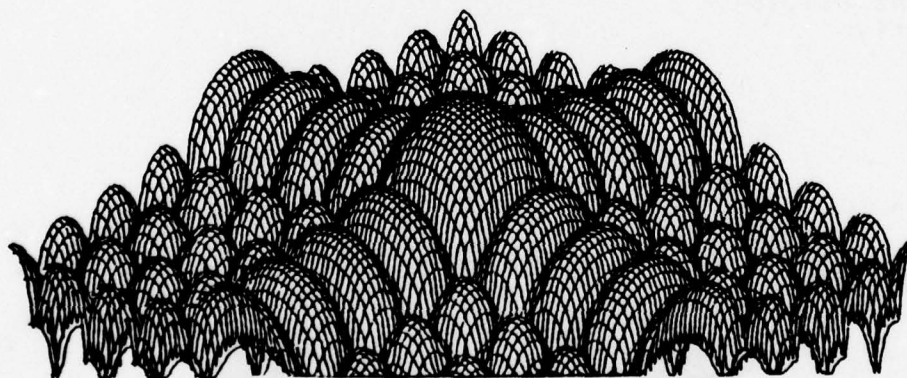




(a) 31x31 separable filter  
 $\beta=1.0$ ,  $\rho=0.21$

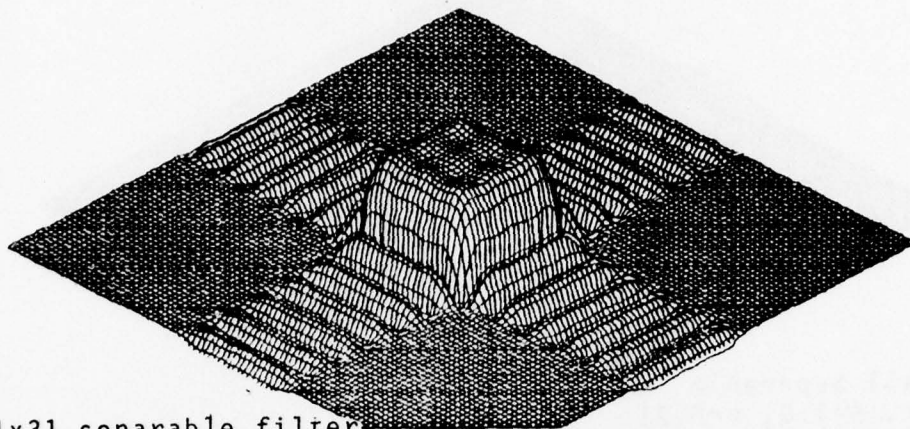


(b) 15x15 separable filter  
 $\beta=1.0$ ,  $\rho=0.14$

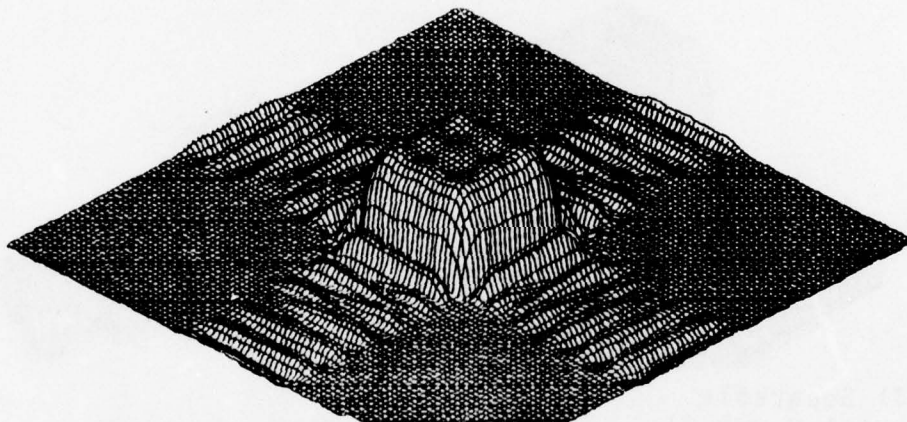


(c) 9x9 separable filter  
 $\beta=1.0$ ,  $\rho=0.05$

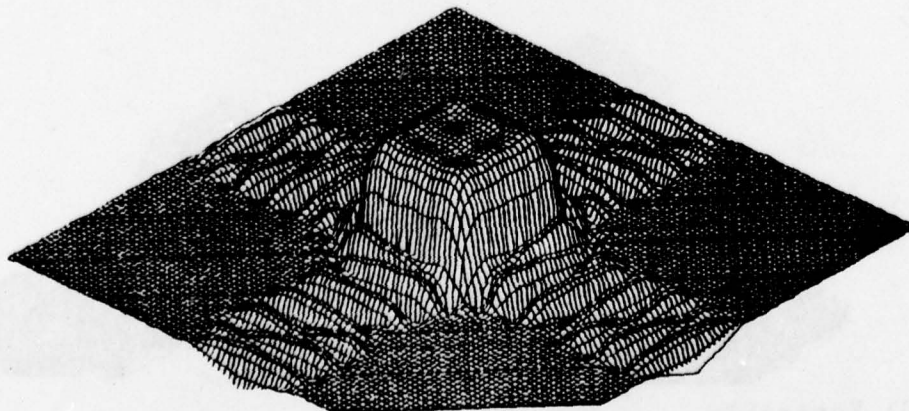
Figure 4.4.3-2  
 Frequency Responses for  
 $N=31 \times 31$ ,  $15 \times 15$ , and  $9 \times 9$   
 Separable FIR Filters Having  
 No Defective Elements



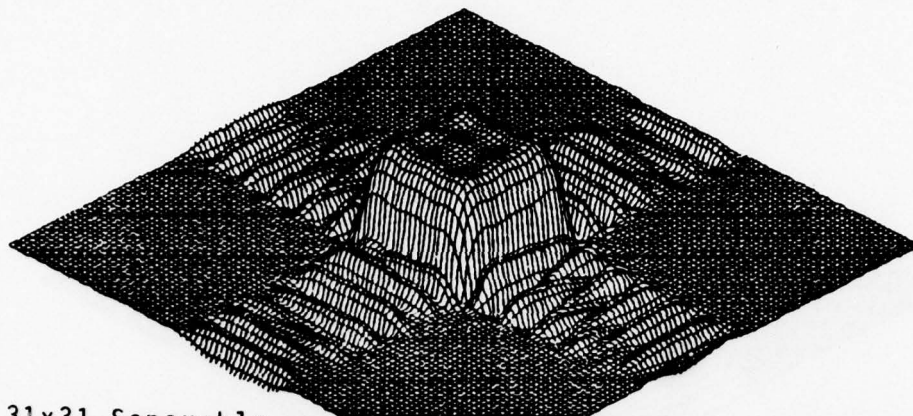
(a) 31x31 separable filter  
 $\beta=1.0$ ,  $\rho=0.21$ , element  
 $(16,16) = 0$



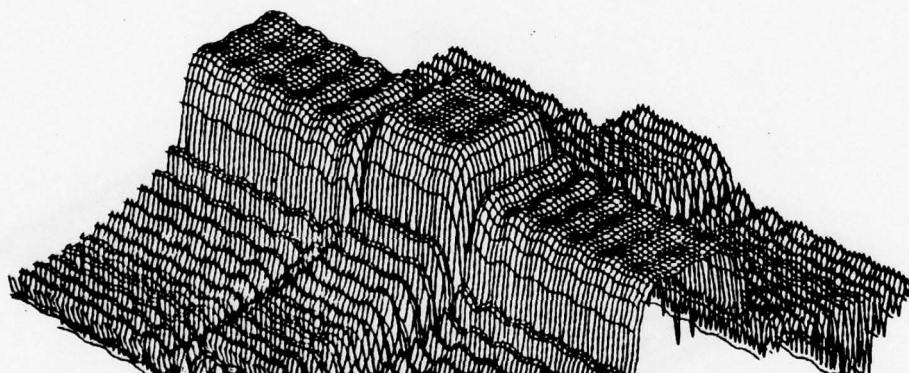
(b) 31x31 separable filter  
 $\beta=1.0$ ,  $\rho=0.21$ , element  
 $(17,17) = 0$



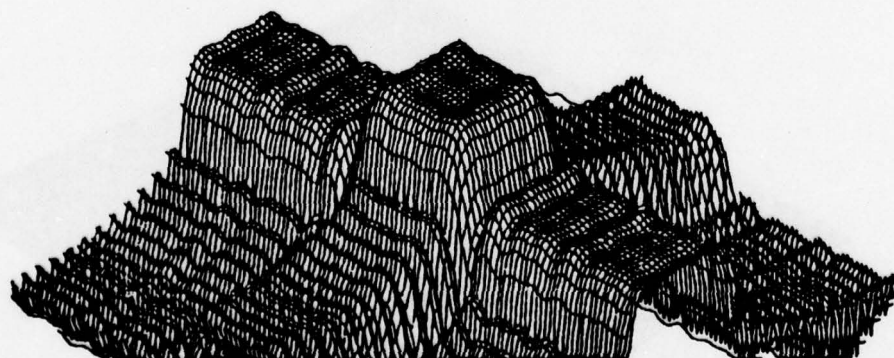
(c) 31x31 separable filter  
 $\beta=1.0$ ,  $\rho=0.21$ , element  
 $(18,18) = 0$



(a) 31x31 Separable  
Filter,  $\beta=1.0$ ,  $\rho=0.21$   
Element (15,18) = 0



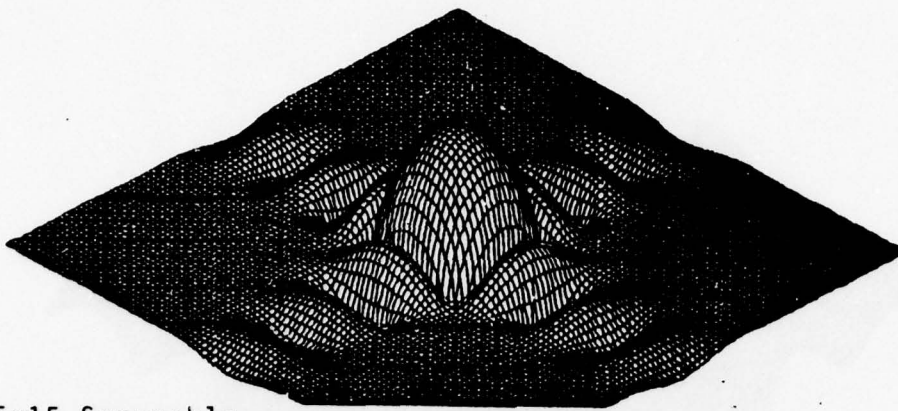
(b) 31x31 Separable  
Filter,  $\beta=1.0$ ,  $\rho=0.21$   
Elements (16,all) = 0



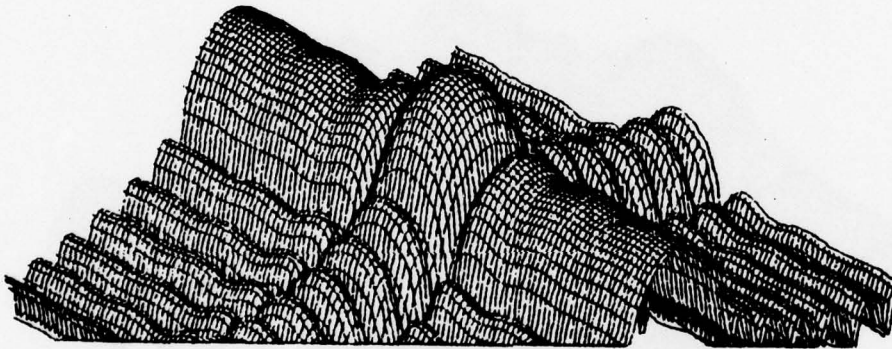
(c) 31x31 Separable  
Filter,  $\beta=1.0$ ,  $\rho=0.21$ ,  
Elements (19, all)=0

Figure 4.4.3-4. Defective Frequency Responses for 31x31  
Separable FIR Filter: PSF Elements (15,18), (16,all),  
and (19, all) Defective

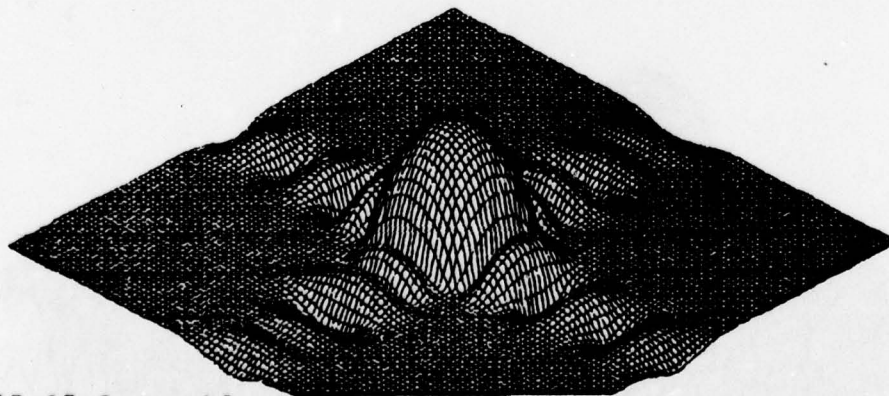




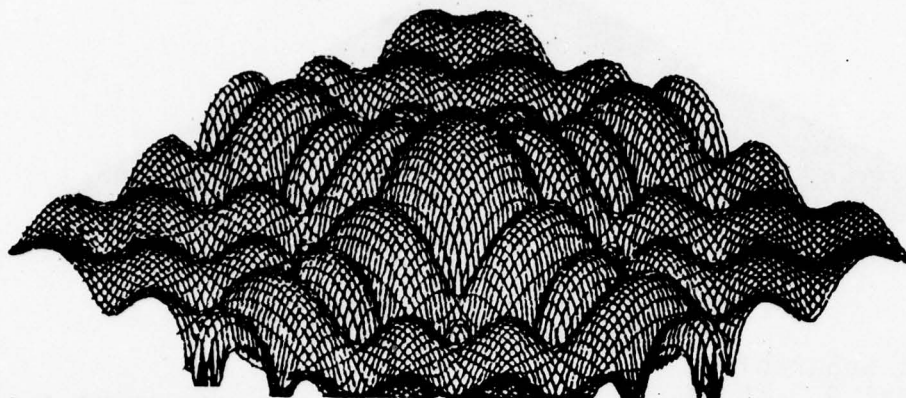
(a) 15x15 Separable  
Filter  $\beta=1.0$ ,  $\rho=0.14$ ,  
Element (8,8)=0



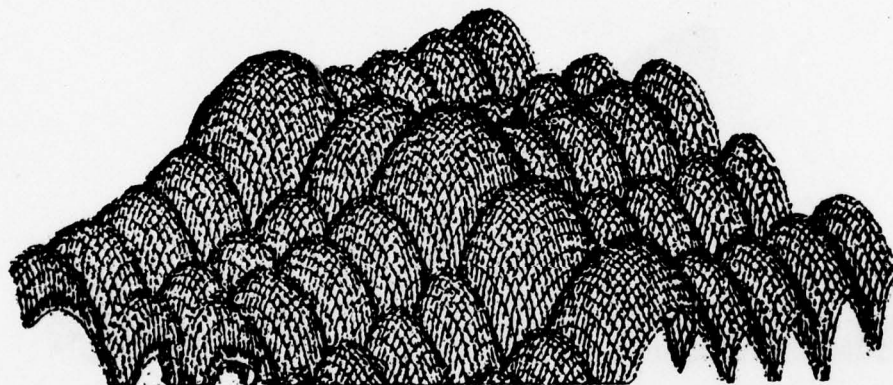
(b) 15x15 Separable  
Filter  $\beta=1.0$ ,  $\rho=0.14$ ,  
Element (8,all)=0



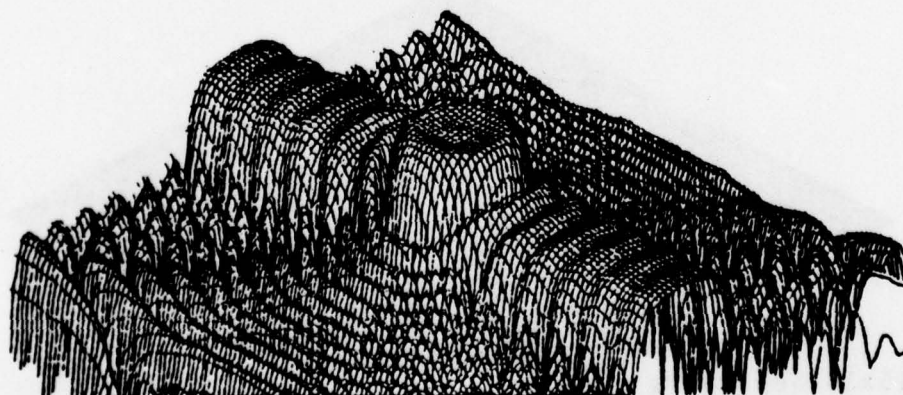
(c) 15x15 Separable  
Filter  $\beta=1.0$ ,  $\rho=0.14$ ,  
Element (10,10)=0



(a) 9x9 Separable  
Filter  $\rho=0.05$ ,  $E$   
Element (5,5)=0



(b) 9x9 Separable  
Filter  $\rho=0.05$ ,  
Element (5,all)=0



(c) 31x31 Nonseparable  
Filter,  $\rho=0.21$ ,  $\beta=1.0$ ,  
Element (16,all)=0

Figure 4.4.3-6 Defective Frequency Responses for 9x9  
FIR Separable Filters With PSF Elements (5,5) and (5,all)  
Defective and for 31x31 FIR Nonseparable Filter with  
PSF Element (16,all) Defective

The remainder of the figures deal with separable filters (with one exception) having Kaiser parameter  $\beta=1.0$  and dimensions  $31 \times 31$ ,  $15 \times 15$ , and  $9 \times 9$ . These filters were designed so that their first stopband null fell at the quarter-bandwidth point. The fact that lower order filters have wider transition band for all other parameters constant means that the value of  $\rho$  will decrease for smaller values of  $N$ . For the filters shown in the following figures, the value of  $\rho$  is 0.21, 0.14, and 0.05 for  $N=31$ , 15, and 9, respectively. Although subjective evaluation of the decimated/interpolated images showed that larger values of  $\rho$  are appropriate for a decimation/interpolation factor of four, results shown in these figures indicate the general nature of the effects of defective filter (or sensor) elements upon filter transfer functions.

Figure 4.4.3-2 shows the frequency transfer functions for separable filters having  $N=31$ , 15, and 9 and no defective elements. These results serve as the basis for comparison of the defective transfer functions.

Figure 4.4.3-3 shows the defective frequency responses corresponding to defective elements located at positions (16,16), (17,17), and (18,18) for the  $31 \times 31$  FIR filter. In all three cases the stopband attenuation in the diagonal frequency direction (halfway between the horizontal and vertical frequency axial directions) is degraded but does not decrease to less than the along-axis value. For the element (16,16)=0 case the symmetry of the frequency response is preserved; setting the center element (16,16) of the point spread function (PSF) to zero does not disturb the symmetry of the PSF about its center element. The defective response of the FIR filter having element (16,16)=0 therefore still has linear phase. The misshapen along-axis stopband sidelobes evident in Figure 4.4.3-3(b) and 4.4.3-3(c)



are a result of the destruction of the symmetry of the PSF that occurs whenever an element other than the center one is set to zero.

Figure 4.4.3-4(a) shows the response for the case in which element (15,18) is defective; this response is not radically different from the previous ones which had defective elements along the diagonal of the PSF, viz., (16,16), (17,17), and (18,18). The (b) and (c) portions of Figure 4.4.3-4 show the frequency responses for the cases in which elements (15, all) and (19, all) are defective, respectively. The "all" notation indicates that the element values of an entire row or column of the PSF matrix are zero. As might be expected, setting an entire row of elements to zero has a profound effect upon the frequency response.

Furthermore, the closer the defective row is to the center of the PSF the greater the degradation of the frequency response. The remarkable fact is that the frequency response degradation resulting from a defective PSF row more than two rows away from the center row is apparently not devastating, judging from Figure 4.4.3-2(a) and Figure 4.4.3-4(b) and (c).

Figure 4.4.3-5 shows the effect upon the frequency response of a 15 x 15 FIR filter having one element or one row of the PSF equal to zero. As in the 31 x 31 case, losing a single FIR element, even the center one, is not catastrophic, whereas losing the center row of the PSF degrades filter performance significantly along one frequency axis. Note the slight change in stopband ripple structure between the (8,8) case and the (10,10) case, (a) and (c) of Figure 4.4.3-5, respectively.

The 9 x 9 separable FIR filter frequency responses corresponding to defective PSF elements (5,5) and (5, all) are shown in the (a) and (b) portions of Figure 4.4.3-6, respectively. Figure 4.4.3-6(c) shows that the

previously observed behavior regarding defective rows in separable PSF's is also characteristic of nonseparable filter PSF's as well. Again, losing a single PSF element does not significantly impair filter response, whereas losing the center row of the PSF causes significant degradation in the frequency response along one axis.

The results presented in this section have indicated that, generally speaking, losing one element of the  $\rho=.25$  two-dimensional PSF weighting matrix does not significantly degrade FIR filter performance for  $31 \times 31$ ,  $15 \times 15$ , and  $9 \times 9$  filters. However, if the center row of the PSF weighting matrix is lost, filter performance is degraded to such an extent that it is probably unacceptable. For the  $\rho=.5$  half-bandwidth filter, losing the center element of a  $31 \times 31$  filter rendered the filter response useless.

## 5.0 IMAGE BANDWIDTH COMPRESSION

This paragraph begins with an overview of the nature of image redundancy and its relation to source coding techniques. This is followed by a description of image reduction (resolution reduction) via decimation and interpolation which is the basis of the REARCS concept. The theory of operation of the several candidate image compression techniques is explained and a performance versus implementation complexity matrix is given, assuming conventional single element raster scanning.

The relationship between implementation complexity and the nature of the scanning process is alluded to and the stage is set for the further investigation of this relationship in the following Section 6.0, Candidate Systems.

This paragraph is mainly intended to serve as a tutorial overview of image bandwidth compression techniques. The critical issues concerning bandwidth compression implementation complexity tradeoffs, the system impact of the REARCS requirement, and the concomitant two-dimensional filtering requirement assuming array scanning are addressed in more depth in paragraph 6.0, Candidate Systems.

### 5.1 Introduction to the Nature of Redundancy in Imagery and Its Relation to Source Coding

All image signals may be regarded as being the sum of two parts: a predictable or redundant part which contains no information and therefore need not be transmitted, and an unpredictable or nonredundant part called the innovation which contains all the information regarding the image. In general, the redundancy of the image is of two kinds. The first kind of redundancy is evidenced by correlation among spatially adjacent picture elements (pixels), or, in the



two-dimensional spatial frequency domain, by nonuniform power spectral density. The adjacent pixels of an image characterized by a uniform power spectral density are uncorrelated, as may be easily shown using the Nyquist Sampling Theorem. The second kind of redundancy is evidenced by the nonuniform probability density function of the image pixel values. This simply means that for signals which have been quantized in amplitude, some amplitude levels occur more frequently than others. This kind of redundancy can be reduced by assigning shorter channel code words to the more frequently occurring amplitude levels via a statistical coding procedure.

Source coding techniques fall into two broad categories: entropy preserving and nonentropy preserving. Entropy preserving techniques are those for which the reconstructed (decoded) image is exactly the same as the original source image, i.e., there is no degradation of image quality. The average number of bits/pixel at the output of an entropy preserving coder can approach the entropy of the source for sufficiently complex coding techniques; note that the entropy of the source is a measure of the information content of the innovation signal and that the predictable or redundant signal has zero entropy. The average bandwidth compression ratio for such coders is image source dependent, and, moreover, varies within an image depending upon the activity of the local part of the image that is being coded; this implies that a channel buffer is required if the channel is to be fully utilized. Nonentropy preserving techniques can obtain large bandwidth compression ratios at the expense of the quality of the reconstructed image. For these techniques, a replica of the innovation signal is not available at the receiver since the channel bit rate is insufficient to match the entropy of the source. Shannon's rate distortion theory mathematically

describes the bound for the trade-off between the fidelity (mean squared error criterion) of the reconstructed image and the image bandwidth compression ratio for any nonentropy preserving source coding technique. This permits the performance of any source coding technique to be evaluated in terms of how closely its rate distortion characteristic approaches the Shannon rate distortion bound.

For nonadaptive, nonentropy preserving source coders, the bandwidth compression ratio is constant regardless of the activity of the image area being encoded; this implies that the quality of the reconstructed image is degraded more in the high activity regions of the image. For the adaptive source coder this is not necessarily the case.

One method of reducing the number of bits that are used to represent a source image is to reduce the number of pixels in the source image via decimation and interpolation. This is a nonentropy preserving technique. The penalty for reducing the required image transmission bandwidth in this way is reduced resolution in the reconstructed image and the possible introduction of spurious detail (called aliasing) into the reconstructed image. The degree of degradation due to aliasing depends upon the sophistication of the two-dimensional decimation and interpolation filters. This aspect of bandwidth reduction via decimation/interpolation was discussed in detail in Section 4.0.

The decimation/interpolation technique can be very profitably combined with conventional source coding techniques in order to obtain very large image bandwidth compression ratios. This approach to image bandwidth compression, called the Redundant Area Coding System (REARCS), was developed by RADC and Harris under contract to RADC in connection with the Experimental Image Compression System (EICS). The primary feature of this philosophy of image bandwidth

compression is that it takes full advantage of the fact that for typical military imagery certain areas of the image (called nonredundant areas) are more important than the remainder of the image (called redundant area). Typically, the nonredundant areas of an image contain objects that are of high military interest, whereas the redundant areas may include rivers, fields, or other geographical features which are used primarily for orientation of the image. Imagery transmission systems utilizing the REARCS concept require that the nonredundant areas be designated by the operator prior to transmission of the image. This is accomplished by dividing the source image into 1-inch squares using a transparent grid template and identifying all 1-inch square blocks which are considered nonredundant areas (refer to Figure 4.1.1). Once the nonredundant areas have been identified, a desired resolution and source coding technique is selected for both the nonredundant and redundant areas of the image. Typically, moderate or high resolution and entropy preserving source coding is required for the nonredundant areas (implying moderate bandwidth compression) whereas low resolution and nonentropy preserving source coding is permissible in the redundant areas (implying very high bandwidth compression). Thus, the REARCS technique permits the interesting portions of an image to be transmitted with high quality while transmitting the remainder of the image at a lower quality. The result is that the objectives of a military image transmission system are met while substantially decreasing the required transmission time per image.

The issues involved with image reduction via decimation/interpolation were discussed in detail in paragraph 4.0.



## 5.2 Image Bandwidth Compression Techniques

The theory of operation of the two-dimensional orthogonal transformation technique, the hybrid technique, multiple tap Delta modulation, statistical coding, DPCM, and MAPS is presented in this section. The rate-distortion coding performance for various source coding algorithms [20] is compared to the Shannon rate distortion bound for bandlimited signals in Figure 5.2.

### 5.2.1 Two Dimensional Orthogonal Transformation Techniques

A gray-shade image can be modelled by specifying its brightness  $u(x,y)$  at each spatial coordinate  $(x,y)$ . If the resulting signal is band-limited, it may be represented by a matrix  $P$  of  $n \times n$  discrete picture elements or pixels. In standard pulse code modulation, each of these pixels is quantized into one of  $2^m$  different levels and each level may then be represented by an  $m$ -bit number. It has been shown [14, 15] that if this matrix  $P$  is transformed by certain linear reversible transformations before quantization, fewer bits are needed to represent the image for a given fidelity than with pulse code modulation. This transformation operation is usually represented by

$$F = APA^T \quad (5-1)$$

where  $A$  is the transform matrix and  $T$  denotes transpose. In this equation  $A$  operates on the columns of  $P$  and  $A^T$  operates on its rows. All terms in this equation are  $n \times n$  matrices;  $F$  is the transformed image. The final step in the coding procedure is to quantize the elements in  $F$  to form  $F_q$ .

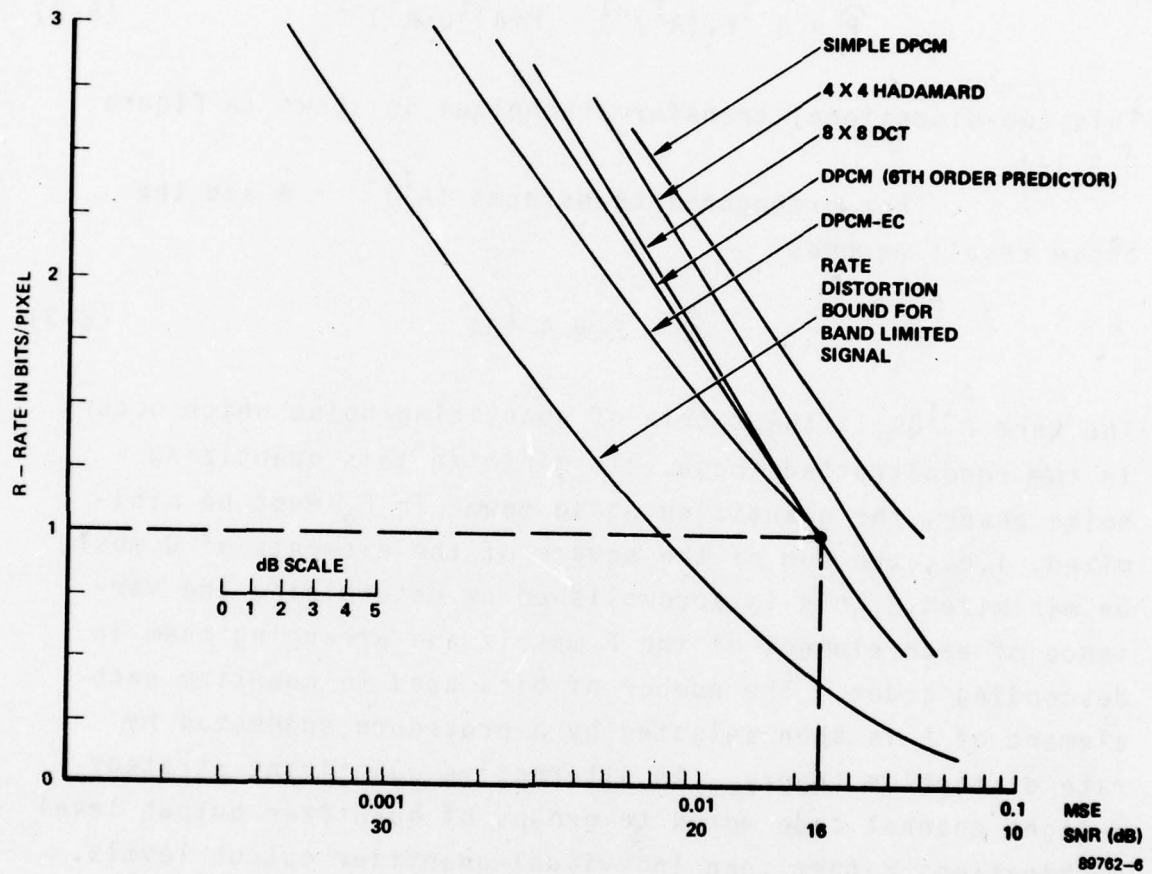


Figure 5.2  
Coder Performance for Unweighted Mean Square Error

Each element in  $F_q$  is then represented by a binary number and  $F_q = F + Q$  where  $Q$  is an  $n \times n$  matrix of the quantizing noise. The reconstructed image  $\hat{P}$  is formed by a linear inverse transformation of  $F_q$ :

$$\hat{P} = A^{-1}F_1(A^T)^{-1} = P + A^{-1}Q(A^T)^{-1} \quad (5-2)$$

This two-dimensional transform technique is shown in Figure 5.2.1-1.

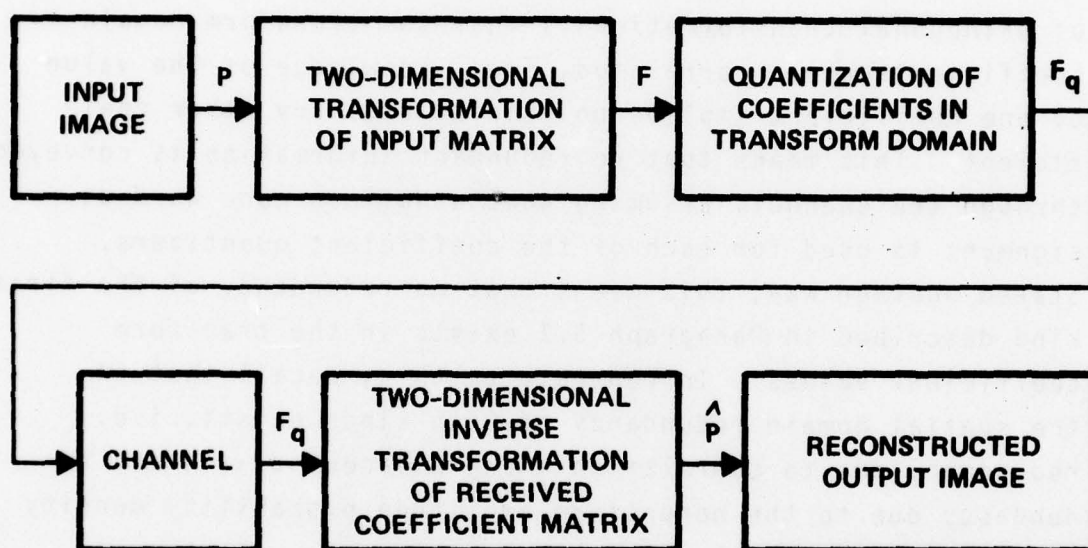
For orthogonal transforms  $(A^T)^{-1} = A$  and the above result becomes

$$\hat{P} = P + A^{-1}QA \quad (5-3)$$

The term  $A^{-1}QA$  is the matrix of quantizing noise which occurs in the reconstructed image. To minimize this quantizing noise power, the quantizing noise power in  $F_q$  must be minimized, i.e., the sum of the square of the elements of  $Q$  must be minimized. This is accomplished by determining the variance of each element of the  $F$  matrix and arranging them in descending order. The number of bits used to quantize each element of  $F$  is then selected by a procedure suggested by rate distortion theory. An alternative quantizing strategy assigns channel code words to groups of quantizer output level combinations rather than individual quantizer output levels.

The two-dimensional transform source coding technique accomplishes bandwidth compression by virtue of two properties of orthogonal transformations. The first property is that the image energy is highly compacted into one corner of the transform domain, i.e., low order transform coefficients have large values and high order transform coefficients have small values. In contrast to this, the



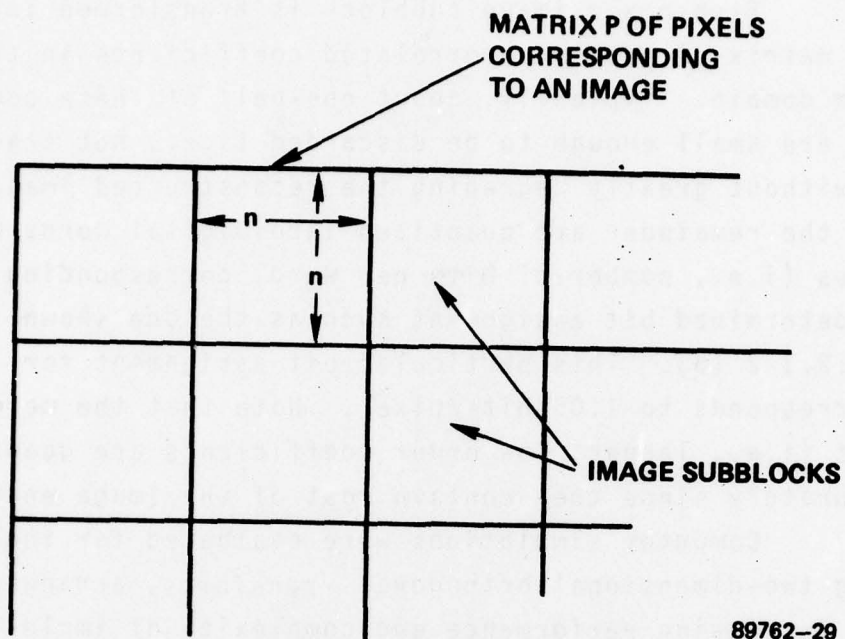


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Figure 5.2.1-1  
Two-Dimensional Transform Technique

image energy (the square of the pixel values) is more uniformly distributed in the spatial domain. This energy compaction property implies that the high order, negligibly small coefficients need not be transmitted at all and that the resulting increase in the number of available bits can then be used to better specify (i.e., more precisely quantize) the remaining low order coefficients. The second property of orthogonal transformation is that the transform domain coefficients are uncorrelated, i.e., knowledge of the value of one coefficient implies no knowledge of any other coefficient. This means that no redundant information is conveyed through the channel, assuming that a Huffman code word assignment is used for each of the coefficient quantizers. Stated another way, this means that no redundancy of the first kind described in Paragraph 5.1 exists in the transform coefficient values. In contrast to this, recall that in the spatial domain redundancy of both kinds exists, i.e., redundancy due to correlation among adjacent pixels and redundancy due to the nonuniform amplitude probability density function of the pixel values.

In any practical two-dimensional transform bandwidth compression technique, the dimension of the input matrix of pixels (subimages or image subblocks) must be limited to a fairly small value such as 4, 8, 16, or possibly 32 in order to maintain reasonable memory requirements (note that input pixel values, transform coefficient values, and transformation matrix values must be stored). Fortunately, statistical influences in typical images only extend over a relatively small number of pixels. This assertion is supported by the fact that the output image SNR increases by a negligible amount in going from 8 x 8 transform size to 16 x 16 transform size [16]. The relationship between image subblocks of size  $n \times n$  and the input image pixel matrix is shown in Figure 5.2.1-2(a).



89762-29

(a) Relationship of Image Subblocks to the Matrix P of Pixels Representing the Image

	1	2	3	4	5	6	7	8
1	6	4	3	2	2	2	2	
2	4	3	2	2	2	2		
3	3	2	2	2	2			
4	2	2	2	2				
5	2	2	2					
6	2	2						
7	2							
8								

NOT TRANSMITTED

89762-30

(b) Typical Bit Assignment in Transform Domain

Figure 5.2.1-2  
Aspects of Two-Dimensional Transform Coding



Each  $n \times n$  image subblock is transformed into an  $n \times n$  matrix of nearly uncorrelated coefficients in the transform domain. Typically, about one-half of these coefficients are small enough to be discarded (i.e., not transmitted) without greatly degrading the reconstructed image quality; the remainder are quantized into digital words with accuracies (i.e., number of bits per word) corresponding to a predetermined bit assignment such as the one shown in Figure 5.2.1-2 (b). This particular bit assignment for  $n = 8$  corresponds to 1.05 bits/pixel. Note that the more important (i.e., larger) low order coefficients are quantized more accurately since they contain most of the image energy.

Computer simulations were evaluated for the following two-dimensional orthogonal transforms, arranged in order of decreasing performance and complexity of implementation.

- Discrete Cosine Transform (DCT)
- Hadamard Transform
- Haar Transform

The results of these simulations are shown in Figure 5.2.1-3.

The implementation complexity of each of these techniques is dictated by the average number of multiplications and additions required per pixel, where multiplications are much more time consuming than additions or subtractions. The DCT matrix of transformation contains  $(n^2 - n)$  irrational numbers implying  $(n^2 - n)$  multiplications and  $n^2$  additions (ignoring normalization), whereas the Hadamard transformation matrix contains only  $\pm 1$ 's, implying that  $n^2$  additions or subtractions are required. The Haar transformation matrix is sparsely populated, i.e., it contains slightly fewer than  $n^2/2$  zeros, and the non-zero entries are  $\pm 1$ , so that approximately  $n^2/2$  additions or subtractions are required.



(a) 8x8 Cosine, 2 bits/pixel



(b) 8x8 Hadamard,  
2 bits/pixel



(c) 8x8 Haar 2 bits/pixel



(d) Original  
4 bits/pixel

Figure 5.2.1-3  
Computer Simulation Results for  
Three 8x8 Orthogonal Transforms at 2 bits/pixel.

### 5.2.2 Hybrid Coding Technique

The hybrid source coding technique utilizes a one-dimensional transform of length  $n$  in one spatial dimension, followed by Differential Pulse Code Modulation (DPCM) of the transform coefficients in the other spatial dimension as shown in Figure 5.2.2-1 (a). These sequential one-dimensional operations attempt to decorrelate the image data in both spatial dimensions. The one-dimensional transform compacts the image energy into the low-order coefficients and decorrelates the coefficients. The bandwidth compression implications of these two properties of transformation were discussed in Paragraph 5.2.1. The mathematical formulation of the one-dimensional transform is

$$\bar{y} = \bar{x} A \quad (5-4)$$

where

$\bar{x}$  = row vector of input pixel values; dimension  
1 x n

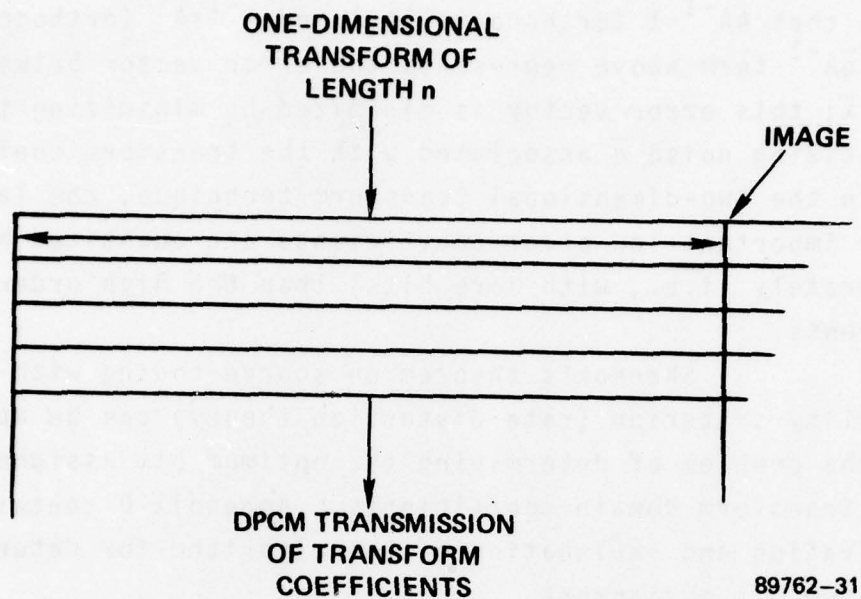
$\bar{y}$  = row vector of transform coefficient values;  
dimension 1 x n

$A$  = matrix of transformation; dimension n x n

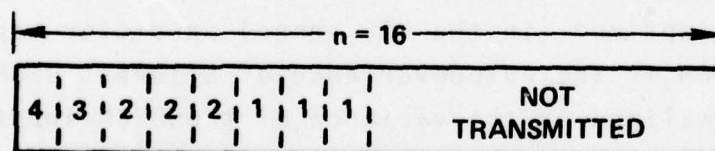
The inverse one-dimensional transformation of the quantized coefficient row vector  $\bar{y}_q = \bar{y} + \bar{q}$  ( $\bar{q}$  is the vector of quantizing noise) is given by

$$\begin{aligned} \hat{\bar{x}} &= \bar{y}_q A^{-1} \\ &= (\bar{y} + \bar{q}) A^{-1} = \bar{y} A^{-1} + \bar{q} A^{-1} \\ &= \bar{x} A A^{-1} + \bar{q} A^{-1} = \bar{x} + \bar{q} A^{-1} \end{aligned} \quad (5-5)$$





Arrangement of One-Dimensional Transform Blocks Within  
A Source Image for Hybrid Coding Technique  
(a)



Bit Assignment for One-Dimensional  
Transform/DPCM Hybrid Coder  
(b)

Figure 5.2.2-1  
Hybrid Coding

Note that  $AA^{-1}=1$  (orthonormality) and  $A^{-1}=A^T$  (orthogonality). The  $\bar{q}A^{-1}$  term above represents the error vector between  $\hat{\bar{x}}$  and  $\bar{x}$ ; this error vector is minimized by minimizing the quantizing noise  $\bar{q}$  associated with the transform coefficients. As in the two-dimensional transform technique, the larger, more important low order coefficients are quantized more accurately (i.e., with more bits) than the high order coefficients.

Shannon's theorem on source coding with a MSE fidelity criterion (rate distortion theory) can be applied to the problem of determining the optimum bit assignment for the transform domain coefficients. Appendix D contains a derivation and explanation of the algorithm for determining the optimum bit assignment.

A particular bit assignment that corresponds to 1.0 bit/pixel for  $n = 16$  is shown in Figure 5.2.2-1 (b).

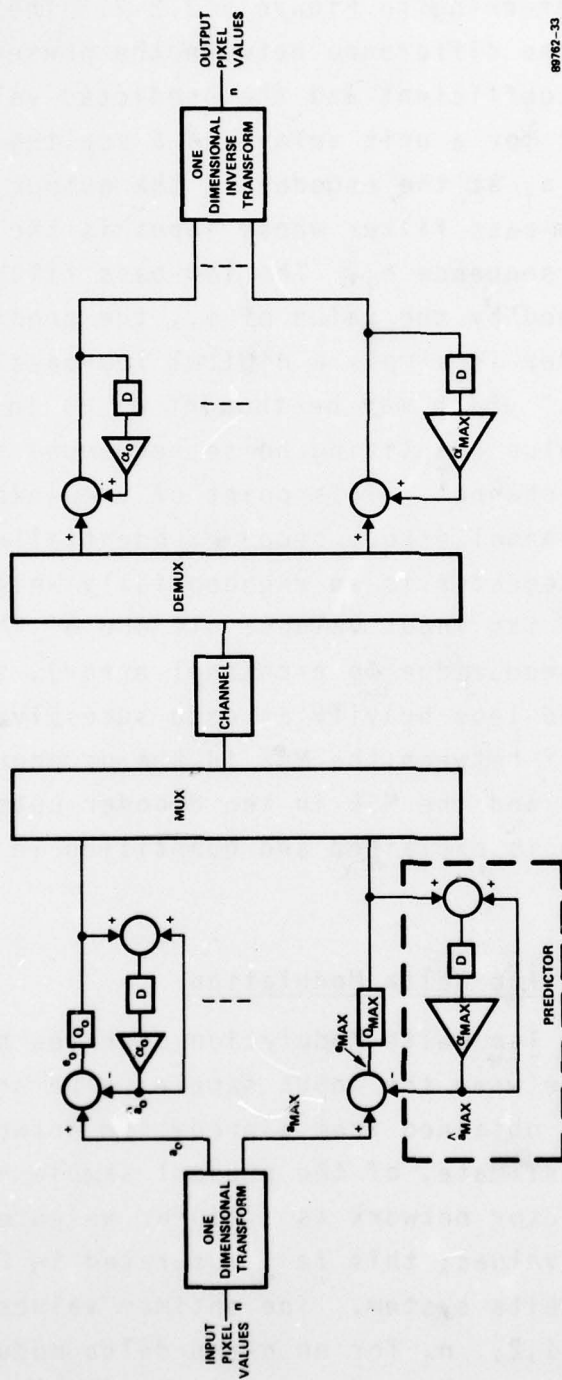
For the special case in which the DPCM predictor tap value is equal to unity, the DPCM operation may be thought of as quantizing and transmitting the difference of coefficient values in adjacent rows. It is shown in Appendix E that the optimum (in the MSE sense) predictor tap value for one-tap DPCM is the autocovariance of adjacent DPCM input values normalized by the variance of the DPCM input values. In the presence of channel errors, the predictor tap values should be decreased somewhat in order to combat the effects of channel errors. Smaller predictor tap values cause the effects of channel errors to die out more quickly. The relationship between predictor tap value and channel noise effects and quantizing noise is derived in Appendix E. Fortunately, the MSE performance of DPCM systems is a rather weak function of the predictor tap value.

Further insight into the operation of DPCM may be obtained by referring to Figure 5.2.2-2. The  $i^{\text{th}}$  quantizer quantizes the difference between the present value  $a_i$  of the  $i^{\text{th}}$  coefficient and the predicted value  $\hat{a}_i$ . Note that except for a unit delay and a scaling factor, the predictor output  $\hat{a}_i$  at the encoder is the output of a recursive digital low-pass filter whose input is the quantized version of the error sequence  $e_i$ . The low-pass filter cutoff frequency is determined by the value of  $\alpha_i$ , the predictor tap value. The decoder is simply a digital low-pass filter, or "leaky integrator," which may be thought of as integrating the error sequence (plus quantizing noise sequence) that is transmitted over the channel. This point of view explains why the effects of channel errors decay exponentially: the output of a leaky integrator is an exponentially weighted past (EWP) average of its input values. If one of the input values is erroneous (due to a channel error), this error will be weighted less heavily at each successive output instant. The tradeoff between the MSE in the decoder output due to channel errors and the MSE in the decoder output due to DPCM source coding is explained and quantified in Appendix E.

### 5.2.3 Multiple Tap Delta Modulation

Multiple Tap Delta Modulation operates by quantizing the difference between the input sample value and an estimate of its value obtained from a predictor network. The predicted value, or estimate, of the present sample value is obtained in the predictor network as a linear weighted sum of selected previous values; this is illustrated in Figure 5.2.3 for a six-tap delta system. The optimum values for the predictor taps  $a_i$ ,  $i=1,2,\dots,n$ , for an  $n$ -tap delta modulation



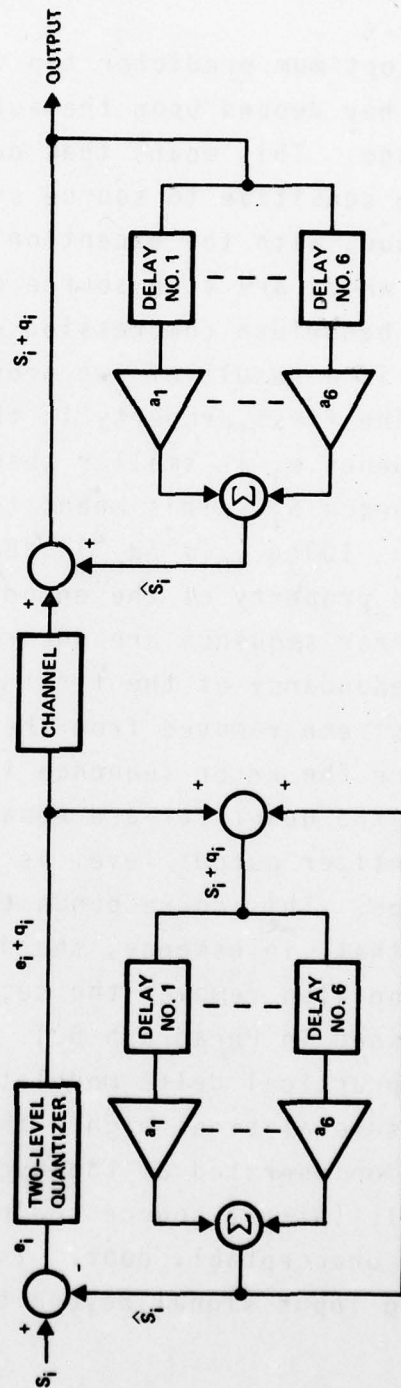


89762-33

Figure 5.2.2-2  
Block Diagram for  
One-Dimensional Orthogonal Transform/DPQM Hybrid Technique



a.



89762-24

b.

Figure 5.2.3  
Six-Tap Delta Modulation

system with linear prediction are the solutions of a system of  $n$  equations in  $n$  unknowns; this result is derived in Appendix E.

The optimum predictor tap values are input data dependent since they depend upon the autocovariance matrix of the source image. This means that delta modulation encoders are more sensitive to source statistics than the transform technique, with the exception of Karhunen-Loeve transformations, which are also source dependent.

The bandwidth compression obtainable using delta modulation is a result of two properties of the source encoding loop. The first property is that the variance  $\sigma_e^2$  of the error sequence  $e_i$  is smaller than the variance  $\sigma^2$  of the input sequence  $S_i$ . This means that the same SNR can be obtained using  $10 \log_{10} (\sigma^2 / \sigma_e^2)$  dB/(6dB/bit) fewer bits. The second property of the encoding loop is that the samples of the error sequence are uncorrelated. This means that the image redundancy of the first kind discussed in Paragraph 5.1 has been removed from the input sequence. In addition, since the error sequence is zero-mean the two output levels of the quantizer are equally likely and (trivially) each quantizer output level is represented by a code word of length one. This corresponds to the simplest possible Huffman code so that, in essence, the delta modulation encoding operation also removes the second kind of image redundancy discussed in Paragraph 5.1.

In practical delta modulation systems the sampling rate is usually several times higher than the Nyquist rate. If delta modulation operated at the Nyquist rate the one bit per Nyquist sample (pixel) source coding performance that results is often unacceptably poor. Increasing the sampling rate of an analog input signal beyond the Nyquist rate



increases the correlation between adjacent samples and, therefore, increases the efficiency of the linear prediction portion of the coder. The penalty for this improved performance is that there are now several channel bits per Nyquist sample since there are several coder samples per Nyquist interval. This is a favorable tradeoff of complexity and performance for some situations, and, in fact, delta modulation out performs DPCM for RC-type signal spectra at low bit/(Nyquist sample) rates [17].

#### 5.2.4 DPCM

Multiple-tap Differential PCM (DPCM) operation is the same as that of multiple-tap delta modulation shown in Figure 5.2.3 except that the simple two level quantizer (comparator with  $\pm k$  volt output) is replaced with a more complex  $n$ -bit or  $N$ -level quantizer [19], where usually  $N=2^n$  in order to fully utilize the bits of the quantizer output codewords. Quantizers of more than two levels are much more complex than comparators and there are performance/complexity tradeoffs among these. For example, nonuniformly spaced quantizing levels yield improved performance for signals having nonuniform probability density functions, but are more complex and expensive to implement than quantizers with uniformly spaced output levels. Optimum nonuniform Laplace quantizers give the best performance in DPCM systems, but optimum uniform Laplace quantizers are easier to implement [27, 28]. In practice, DPCM encoders usually operate at or near the Nyquist sampling rate, whereas delta modulation systems almost invariably operate with a sampling frequency several times higher than the Nyquist rate for reasons explained in Section 5.2.3. DPCM encoding followed by

Huffman coding of the DPCM quantizer outputs gives improved rate-distortion performance, but at the expense of the additional complexity associated with channel buffers and Huffman coders.

#### 5.2.5 Statistical Coding

Statistical coding refers to an entropy preserving technique in which a Huffman code is used to transmit the difference in gray shade of adjacent pixels in the along-scan direction of the digitized image. The block diagram of this coding technique is shown in Figure 5.2.5. The bandwidth compression of statistical coding is a result of the high degree of spatial correlation between adjacent pixels in typical imagery. The amplitude probability density function of the difference of adjacent pixels is very sharply peaked (about zero) as a result, which implies that Huffman coding of the adjacent pixel differences can yield useful bandwidth compression ratios on the order of 2:1 to 3:1, depending upon the image source statistics. The primary advantage of statistical coding is that it is an entropy preserving technique, i.e., the reconstructed image is exactly like the source image, in the absence of channel errors. The disadvantages are three-fold: the Huffman code is source-dependent, the nonconstant encoder output bit rate requires channel buffers at the encoder and decoder, and the effect of channel errors propagates indefinitely in the reconstructed image. The propagation of the effects of channel errors can be controlled by periodic reinitialization with an acceptably small penalty in overhead rate but at the cost of increased complexity.

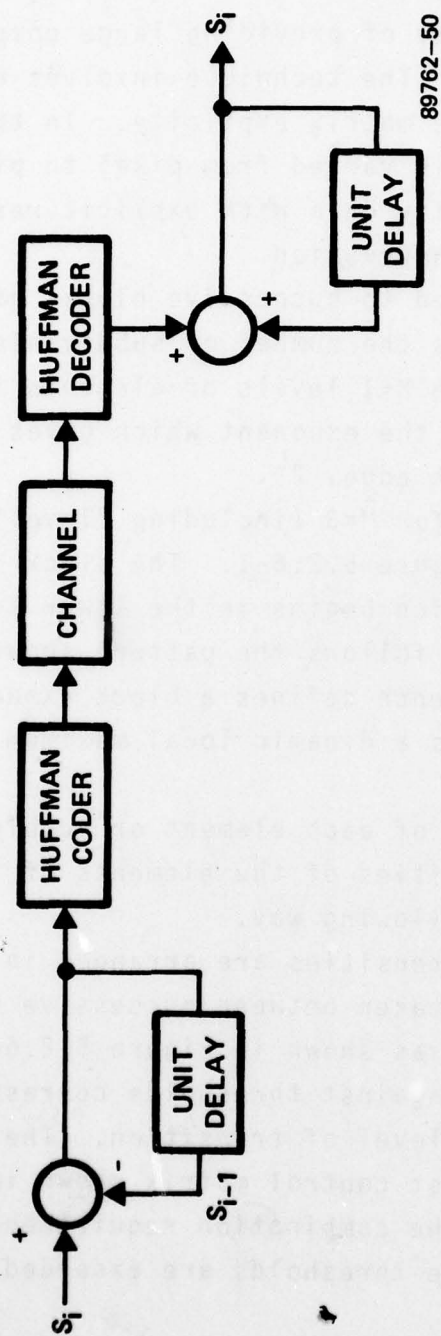


Figure 5.2.5  
Statistical Coding Technique



#### 5.2.6 MAPS Algorithm Description

The Micro Adaptive Picture Sequencing (MAPS) technique has the capability of providing large compression ratios in digital imagery. The technique involves defining the resolution of the image matrix explicitly. In this process, the element size is varied from pixel to pixel. Replacing redundant intensity data with explicit resolution information provides the compression.

MAPS is applied to successive blocks of pixels of size  $2^M \times 2^M$  where  $M$  is the number of subdivisions within the block. Thus, there are  $M+1$  levels of elements in each block. This level is just the exponent which gives the number of pixels along the element edge,  $2^L$ .

The elements for  $M=3$  (including "level" as a subscript) are shown in Figure 5.2.6-1. The block sequence used is an "inverted z" which begins in the lower left corner of each element and follows the pattern shown in Figure 5.2.6-2. This sequence defines a block exhaustion configuration which imposes a dynamic local maximum block size constraint.

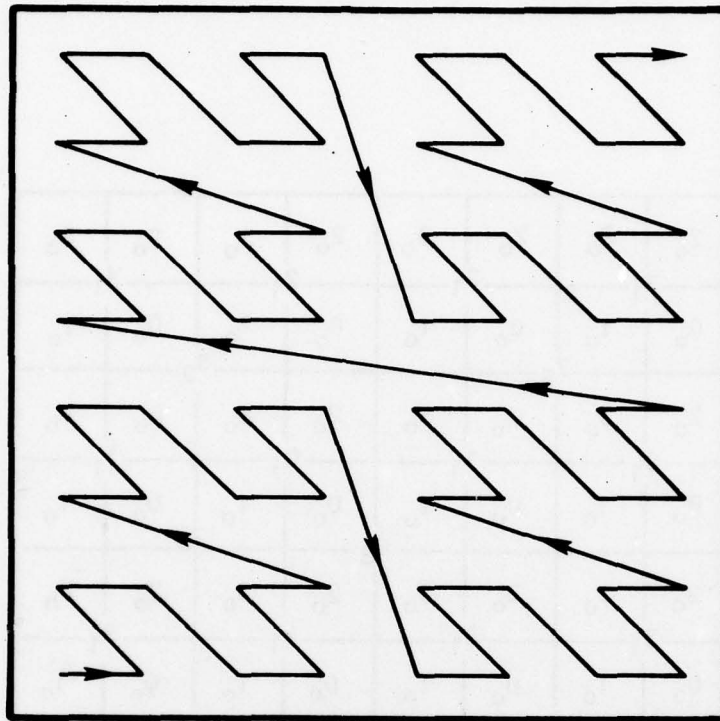
The intensity of each element or subblock is calculated from the intensities of the elements of the next lower level in the following way.

First, the intensities are arranged in increasing order and differences are taken between successive values and between extreme values as shown in Figure 5.2.6-3. These contrasts are then tested against thresholds corresponding to both contrast type and level of transition. The contrast thresholds form the contrast control matrix shown in Figure 5.2.6-4, which specifies the combination requirements for each level. If none of the thresholds are exceeded, the

$2_o$	$3_o$	$2_o$	$3_o$	$2_o$	$3_o$	$2_o$	$3_o$
$2_1$		$3_1$		$2_1$		$3_1$	
$0_o$	$1_o$	$0_o$	$1_o$	$0_o$	$1_o$	$0_o$	$1_o$
$2_2$		$3_2$		$2_2$		$3_2$	
$2_o$	$3_o$	$2_o$	$3_o$	$2_o$	$3_o$	$2_o$	$3_o$
$0_1$		$1_1$		$0_1$		$1_1$	
$0_o$	$1_o$	$0_o$	$1_o$	$0_o$	$1_o$	$0_o$	$1_o$
$2_3$		$3_3$		$2_3$		$3_3$	
$2_o$	$3_o$	$2_o$	$3_o$	$2_o$	$3_o$	$2_o$	$3_o$
$2_1$		$3_1$		$2_1$		$3_1$	
$0_o$	$1_o$	$0_o$	$1_o$	$0_o$	$1_o$	$0_o$	$1_o$
$0_2$		$1_2$		$0_2$		$1_2$	
$2_o$	$3_o$	$2_o$	$3_o$	$2_o$	$3_o$	$2_o$	$3_o$
$0_1$		$1_1$		$0_1$		$1_1$	
$0_o$	$1_o$	$0_o$	$1_o$	$0_o$	$1_o$	$0_o$	$1_o$

10293-5

Figure 5.2.6-1  
Subframe Partition For M=3



10293-7

Figure 5.2.6-2  
"Block" Sequence Path



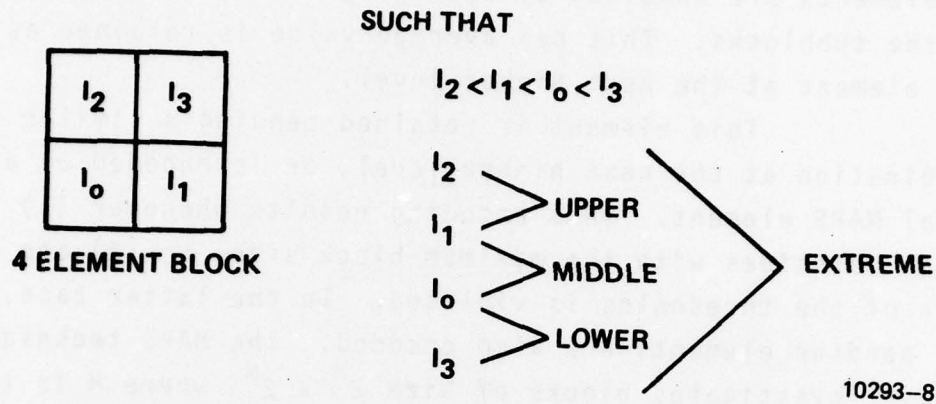


Figure 5.2.6-3. Fidelity Control Intensity/Contrast Definition

LEVEL TRANSITIONS	EXTREME	MIDDLE STEP	LOWER STEP	UPPER STEP
0 → 1	E <sub>1</sub>	M <sub>1</sub>	L <sub>1</sub>	U <sub>1</sub>
1 → 2	E <sub>2</sub>	M <sub>2</sub>	L <sub>2</sub>	U <sub>2</sub>
2 → 3	E <sub>3</sub>	M <sub>3</sub>	L <sub>3</sub>	U <sub>3</sub>
3 → 4	E <sub>4</sub>	M <sub>4</sub>	L <sub>4</sub>	U <sub>4</sub>
4 → 5	E <sub>5</sub>	M <sub>5</sub>	L <sub>5</sub>	U <sub>5</sub>

CONTRASTS

10293-6

Figure 5.2.6-4. Contrast Control Matrix

subelements are combined by averaging the intensity value of the subblocks. This new average value is retained as a new element at the next higher level.

This element is retained pending a similar combination at the next higher level, or is encoded as a final MAPS element. This encoding results whenever (1) the level coincides with the maximum block size, or (2) one or more of the thresholds is violated. In the latter case, all pending elements are also encoded. The MAPS technique always investigates blocks of size  $2^M \times 2^M$  where  $M$  is the maximum number of subdivisions within the block.

The maximum block size is determined by trading off the buffer memory required for temporary storage of the elements and the advantage acquired in achievable compression ratio.

The encoding procedure consists of two parts. The compound codeword is composed of a resolution code which describes the level of the MAPS combination, and an intensity code which defines the local block intensity. The number of bits necessary to describe the resolution is an important factor in choosing maximum block size, since it affects compression ratios directly.

### 5.3

#### EVALUATION OF PERFORMANCE/COMPLEXITY TRADEOFF FOR SOURCE CODING TECHNIQUES

The theory and operation of six source coding algorithms were described in subsections 5.2.1 through 5.2.6.

- Two-dimensional transform coding
- Hybrid coding (one-dimensional transform along rows followed by DPCM transmission of coefficients along columns)

- Six-tap Delta Modulation
- DPCM
- Statistical coding
- MAPS

In evaluating source coding algorithms the following characteristics should be considered:

- Overall subjective quality of reconstructed imagery
- Resolution of reconstructed imagery
- Normalized MSE performance
- Propagation of channel errors in the reconstructed imagery
- Sensitivity to image source statistics
- Ease of implementation -- impacted by
  - Memory requirements
  - Speed requirements
  - Arithmetic precision requirements
  - Synchronization/interface requirements
  - Flexibility requirements
  - REARCS requirements

The results of the computer simulations for some of the candidate coding techniques were used to help construct the trade-off matrix shown in Table 5.3.

Based upon the simulation results and considering the results contained in the wealth of literature regarding image bandwidth compression [16-26] the hybrid coding technique utilizing a one-dimensional Discrete Cosine Transform (DCT) appears to be an appropriate choice to meet the bandwidth compression requirement. It is claimed in references [16, 21, 22 and 23] that the performance of hybrid



TABLE 5.3 Performance/Complexity Matrix for  
Several Source Coding Techniques

CHARACTERISTIC	Two-Dimensional Transform	One-Tap DPCM	Hybrid	Six-Tap Delta Modulation	MAPS	Statistical Coding
Subjective Image Quality	FAIR	FAIR	GOOD	FAIR	GOOD	EXCELLENT
Image Resolution	GOOD	FAIR	GOOD	FAIR	FAIR	EXCELLENT
Normalized MSE Performance	FAIR	FAIR	GOOD	POOR	-	-
Immunity to Channel Error Effects	FAIR	FAIR	FAIR	FAIR	-	*POOR TO MODERATE
Insensitivity to Source Statistics	GOOD	POOR	GOOD	POOR	GOOD	POOR
Implementation Complexity	VERY HIGH	LOW	HIGH	LOW	HIGH	*MODERATE TO HIGH

\*Depending upon channel error control strategy

-Not applicable or unknown

coding with DCT is superior to that of two-dimensional transform coders, and, further, that hybrid coders perform nearly as well as optimal Karhunen-Loeve transform coders.

The performance of the hybrid technique at 1.0 bit/pixel is very good even in channels characterized by  $\text{BER}=10^{-3}$  (worst-case); typical channels should exhibit a nominal  $\text{BER}=10^{-5}$ . A reconstructed hybrid-coded image is shown in Figure 5.3 for  $\text{BER}=10^{-3}$ ; the unprocessed original is also shown there for comparison. The performance versus complexity tradeoff remains to be evaluated under the array scanning assumptions.

The basic source coding algorithms that have been described in this section, with the exception of the two-dimensional transform, hybrid and MAPS techniques, all accept pixel values sequentially, one at a time. The two-dimensional transform, MAPS, and hybrid bandwidth compression algorithms operate on two-dimensional or one-dimensional arrays of image data; for scanning systems which output pixel values sequentially, some storage medium is required in order to accumulate the one or two-dimensional image data blocks required for these three algorithms. For single point scanning systems the storage requirement is such that digital storage must be used; the associated control functions and digital arithmetic functions, combined with the sheer quantity of digital storage required, lead to implementations of high complexity. For array scanning systems, the one- or two-dimensional nature of the sensing and recording techniques is better suited to the block data structure required for hybrid and two-dimensional transformation bandwidth compression algorithms.

In this section we have discussed the concepts involved in image bandwidth compression and described the operation of six source coding algorithms. The performance and

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HARRIS CORP MELBOURNE FL GOVERNMENT SYSTEMS GROUP  
ARRAY SCANNING FOR IMAGE TRANSMISSION.(U)  
JAN 79 W C ADAMS, T R BADER, J J KNAB

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(a) Reconstructed Image  
1.0 bit/pixel,  $BER = 10^{-3}$



(b) Original Source Image  
6 bit/pixel

Figure 5.3  
Original Source Image and Reconstructed Image Using  
the Hybrid Technique, 1.0 bit/pixel  $BER=10^{-3}$

complexity for each of these techniques were ranked and the hybrid technique was judged as giving the best overall performance. It is perhaps not surprising that sophisticated algorithms give better performance than simple ones; it is the purpose of the next paragraph, Candidate Systems, to propose and evaluate the cost effectiveness of a variety of ASITS architectures for several of the bandwidth compression techniques discussed in this section. This task is made all the more difficult because of the REARCS requirement, but the REARCS capability is so absolutely vital to achieving the desired overall bandwidth compression ratios that it can hardly be abandoned. It will be shown in the next section that the implementation complexity of the high performance bandwidth compression schemes, viz., the hybrid, MAPS, and two-dimensional transform coding algorithms, is greatly reduced when full advantage is taken of array scanning techniques.

## 6.0

### CANDIDATE SYSTEM ARCHITECTURES FOR ASITS

In this section we present and evaluate a number of ASITS architectures which are capable of implementing various combinations of desirable ASITS functions, including:

- selectable image bandwidth compression ratios
- selectable resolution
- REARCS, which implies selectable image bandwidth compression ratios and selectable resolution for designated portions of the source image
- two-dimensional filtering in conjunction with REARCS
- image enhancement
- image MTF correction

Architectures which do not implement all of these desirable ASITS functions are presented here for several reasons. First, presenting architectures of varying capabilities shows the tradeoff between system capability and implementation complexity in a way that is easy to understand. Second, this method of presenting system architectures facilitates an easier grasp of the various aspects of the ASITS functions. Finally, this approach provides a shopping list for applications which have less sophisticated or less comprehensive system requirements than those we have arbitrarily imposed for the purposes of providing some structure for the study effort.

Recall from Section 5.0 that the hybrid algorithm was considered to be the most desirable high performance bandwidth compression technique for single point scanning systems when one considers performance versus implementation complexity. We now re-evaluate the implementation complexity



of this and the other bandwidth compression techniques discussed in Section 5.0 assuming an array scanning technique rather than a single-point scanning technique. This re-evaluation of bandwidth compression implementation complexity is integrated with an evaluation of the overall system complexity assuming that array scanning is used for the sensing and recording tasks. The approaches to system implementation presented in this section include various degrees of digital signal processing versus sampled-data analog signal processing. Also, since it was shown in Sections 2.0 and 3.0 that one-dimensional scanning techniques are far easier to implement than two-dimensional scanning techniques, all of the candidate systems presented in this section use one-dimensional scanning in conjunction with analog or digital storage in order to synthesize two-dimensional scanning.

In this section we present candidate systems that may be described as follows:

- nonseparable two-dimensional filtering, one-tap DPCM, single spot recording
- separable two-dimensional filtering, one-tap DPCM, REARCS, single spot recording
- separable two-dimensional filtering, one-tap DPCM, REARCS, single spot recording, MTF correction, REARCS
- separable two-dimensional filtering, one-tap DPCM, multiple-spot recording, MTF correction
- two-dimensional DCT transform/PCM with REARCS, MTF correction, multiple-element recording
- hybrid (1 dimensional DCT/DPCM) with REARCS, MTF-correction, multiple-element recording

- All digital implementation of hybrid DCT/DPCM, no REARCS, no filtering
- MAPS BWC, multiple-element recording
- Entropy-coded multitap DPCM and statistical coding

#### 6.1 NONSEPARABLE TWO-DIMENSIONAL FILTERING, ONE-TAP DPCM, MTF CORRECTION, SINGLE-ELEMENT RECORDING

The system architecture to be presented below is relatively simple and performs only two image processing functions: two-dimensional nonseparable filtering and bandwidth compression via one-tap DPCM, an unsophisticated source coding technique when compared to some of the others. This simple system does serve to illustrate the level of implementation complexity of one of the simpler ASITS schemes, and, in addition, is a convenient vehicle for explaining the implementation of two-dimensional FIR filters using a one-dimensional array sensing technique.

Figure 6.1-1 shows how this system would be implemented using a tapped analog delay line of length  $N^2$  in conjunction with  $N^2$  tap weights where  $N$  is the length of the one-dimensional sensing array. The one-tap DPCM source encoding shown here is one-dimensional in nature and does not take advantage of the vertical redundancy existing within the image; more sophisticated techniques are two-dimensional in nature and will be discussed in later sections. Note that the output of the transmitter portion of the system of Figure 6.1-1 corresponds to a single scan line of the reconstructed image and therefore requires only a single-element recording technique.

Figure 6.1-2 is a graphical representation of a two-dimensional nonseparable transversal (FIR) filter

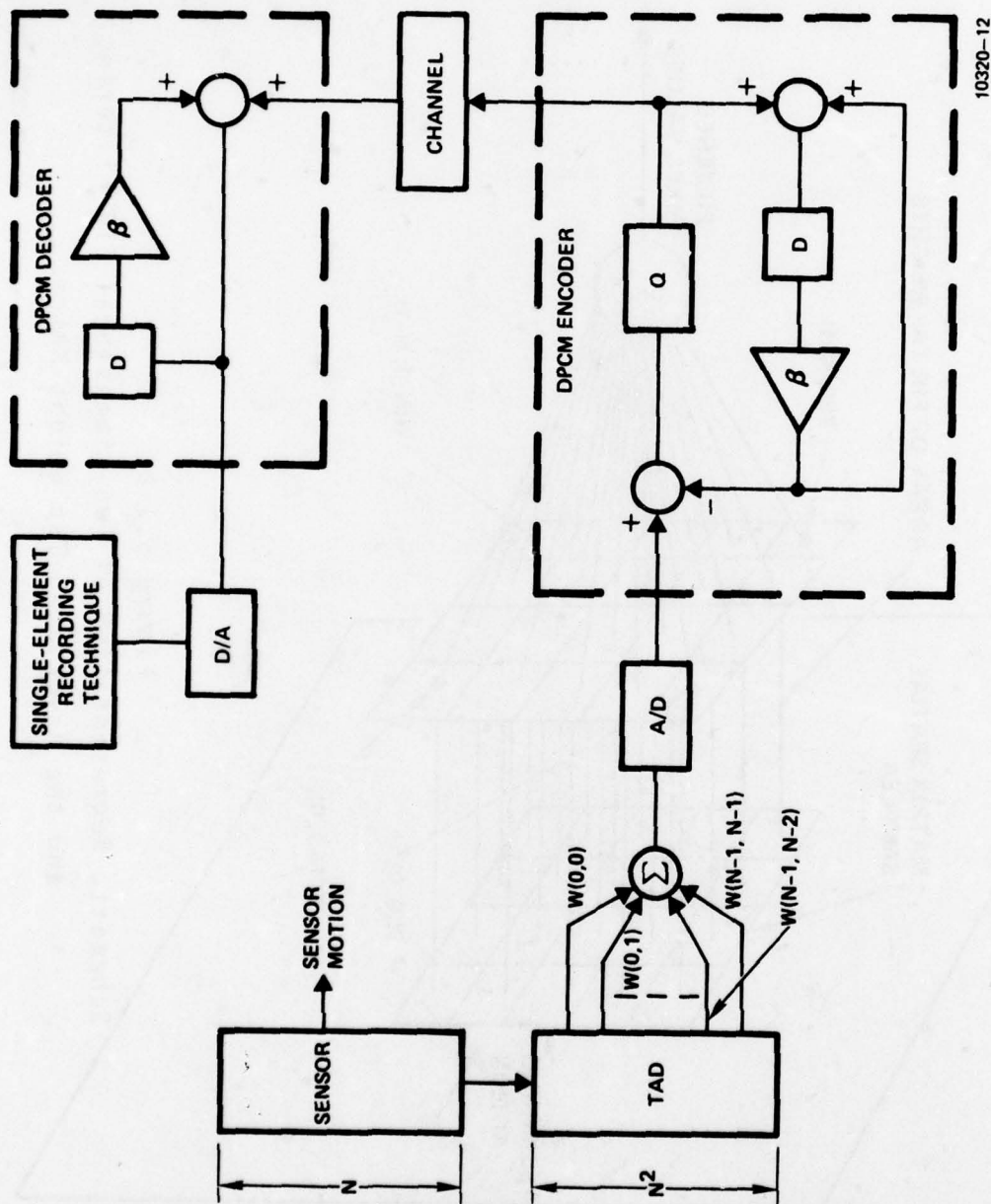


Figure 6.1-1  
Nonseparable Two-Dimensional Filtering, MTF Correction,  
One-Tap DPCM, Single Element Recording



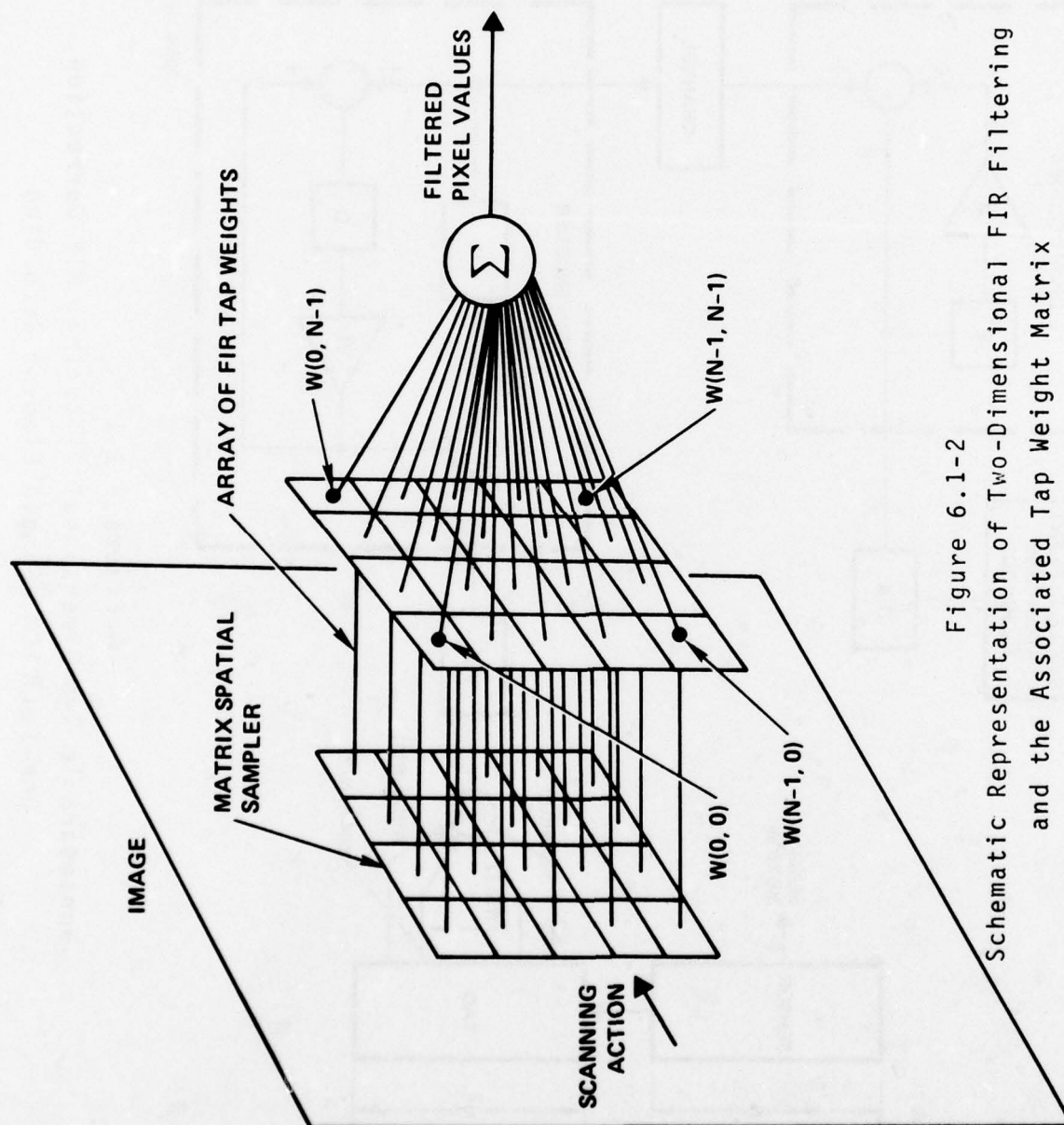


Figure 6.1-2  
Schematic Representation of Two-Dimensional FIR Filtering  
and the Associated Tap Weight Matrix

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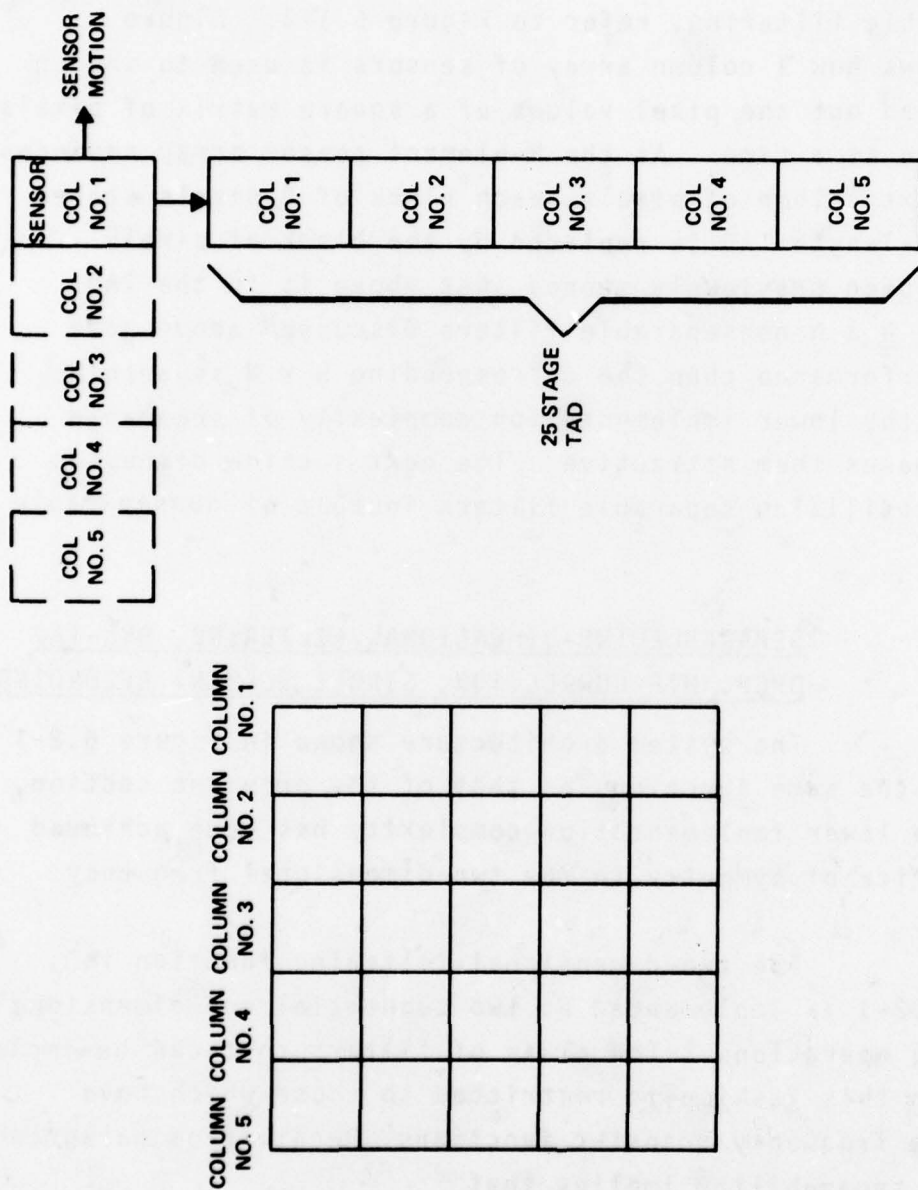
that is driven by a two-dimensional sensor array of the appropriate size. In order to understand how the one-dimensional sensor array of Figure 6.1-1 is used to implement nonseparable filtering, refer to Figure 6.1-3. Figure 6.1-3 shows how a column array of sensors is used to sequentially read out the pixel values of a square matrix of pixels, one column at a time. As the N-element sensor array advances to the next column of pixels, each block of N pixels stored in the  $N^2$ -length TAD is replaced by the block of pixels that had been previously stored just above it in the TAD. While the  $N \times N$  nonseparable filters discussed above have better performance than the corresponding  $N \times N$  separable filters, the lower implementation complexity of separable filters makes them attractive. The next section discusses an ASITS utilizing separable filters instead of nonseparable filters.

## 6.2 SEPARABLE TWO-DIMENSIONAL FILTERING, ONE-TAP DPCM, MTF CORRECTION, SINGLE ELEMENT RECORDING

The system architecture shown in Figure 6.2-1 performs the same functions as that of the previous section, 6.1. The lower implementation complexity has been achieved at the price of symmetry in the two-dimensional frequency response.

The two-dimensional filtering function in Figure 6.2-1 is implemented as two sequential one-dimensional filtering operations. The class of filters that can be implemented in this fashion is restricted to those which have separable frequency transfer functions. Recall from paragraph 4.0 that separability implies that

$$H(f_h, f_v) = H_1(f_h) H_2(f_v) \quad (6-1)$$



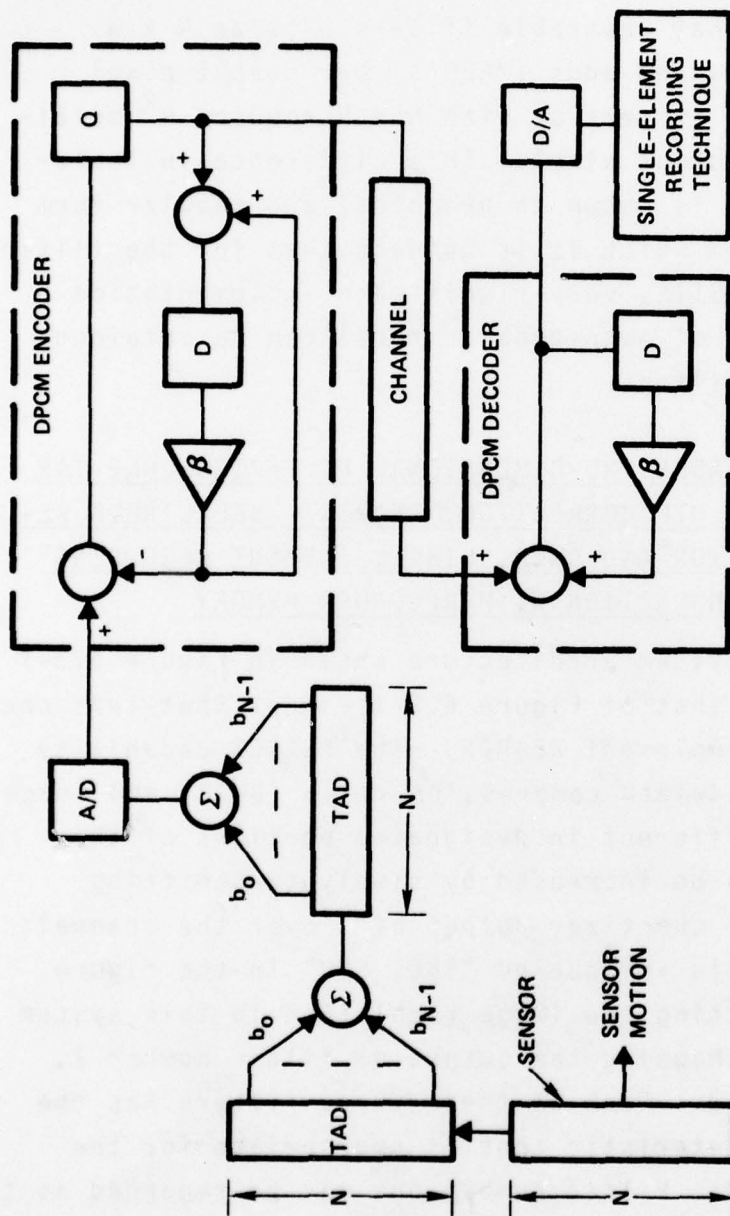
a. 5 X 5 ARRAY OF PIXELS

b.

10320-14

Figure 6.1-3. Implementation of Nonseparable Two-Dimensional Filters Using One-Dimensional Sensor





$b_k$  = FILTER TAP WEIGHTS

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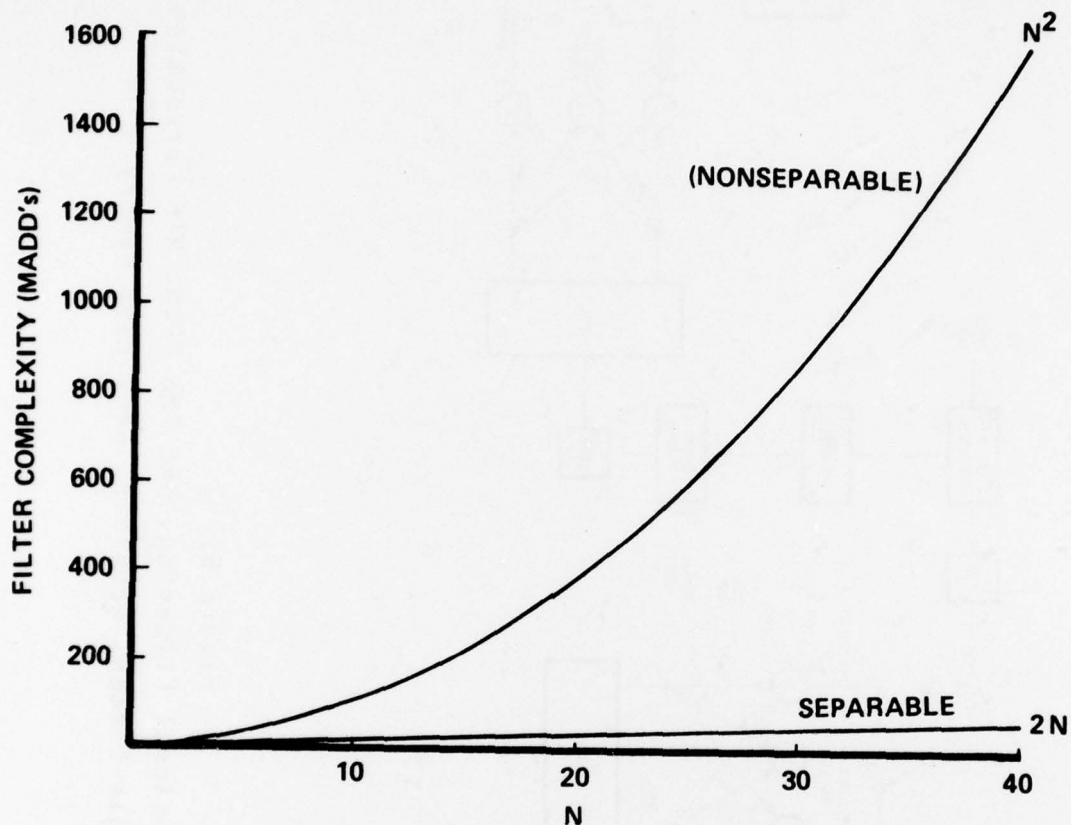
Figure 6.2-1  
Separable Two-Dimensional Filtering, MTF Correction, One-Tap DPCM,  
Single Element Recording

where  $f_h$  and  $f_v$  refer to horizontal spatial frequency and vertical spatial frequency, respectively. The difference in implementation complexity between separable and nonseparable filters increases rapidly with the filter dimension  $N$ . This is due to the fact that separable filters of size  $N \times N$  require  $2N$  multiplies and adds (MADD's) per output pixel whereas nonseparable filters of size  $N \times N$  require  $N^2$  multiplies and adds per output pixel. This difference in implementation complexity is shown in graphical and tabular form in Figure 6.2-2, from which it is evident that for the filter sizes required for ASITS, very significant implementation advantages (an order of magnitude or more) can be obtained by using separable filters.

### 6.3 SEPARABLE TWO-DIMENSIONAL FILTERING, ONE-TAP DPCM, MTF CORRECTION, REARCS - SELECTABLE RESOLUTION AND BWCR, SINGLE ELEMENT RECORDING IN CONJUNCTION WITH RECORDER MEMORY

The system architecture shown in Figure 6.3-1 is quite similar to that of Figure 6.2-1 except that this one has the ability to implement REARCS. The REARCS capability implies that the bandwidth compression ratio (BWCR) and image resolution may be different in designated portions of the image. The BWCR can be increased by simply transmitting fewer of the encoder quantizer output bits over the channel; the command to do this is labeled "BWCR CMD" in the figure.

Selecting the image resolution in this system is accomplished by choosing the output of filter number 1, number 2, or number 3. Each of these three filters has the MTF correction characteristic that is appropriate for the selected resolutions. Filter number one may be regarded as the

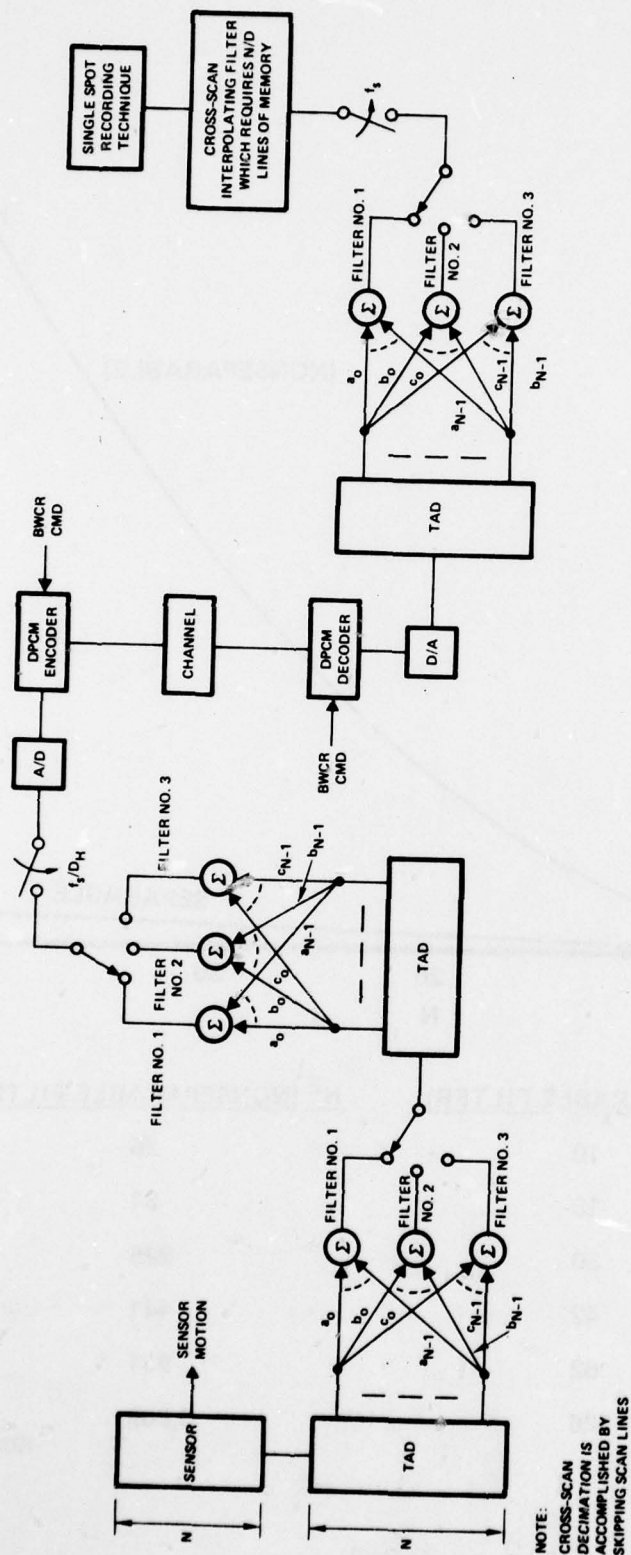


<u>N</u>	<u>2N (SEPARABLE FILTER)</u>	<u>N<sup>2</sup> (NONSEPARABLE FILTER)</u>
5	10	25
9	18	81
15	30	225
21	42	441
31	62	961
63	126	3,969

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Figure 6.2-2  
Filter Complexity of Separable Versus  
Nonseparable Filters  
283



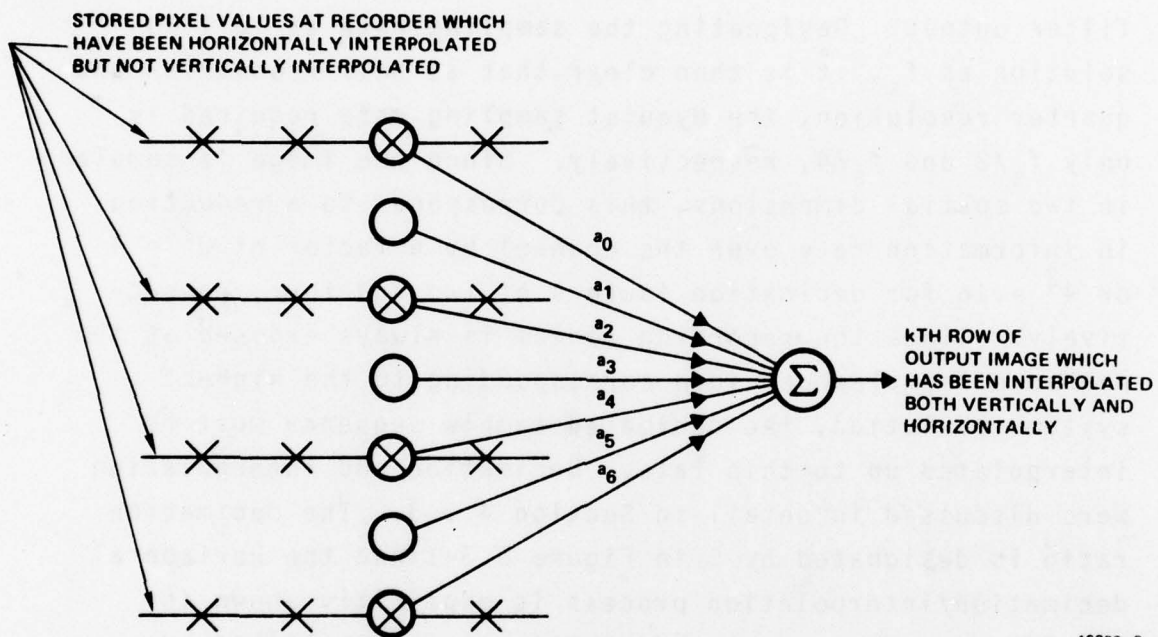


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Figure 6.3-1  
Separable Two-Dimensional Filtering, One-Tap DPCM, MTF Correction,  
REARCS, Single-Element Recording Requiring Memory

full resolution MTF correction filter, whereas filters two and three perform the MTF correction and lowpass filtering appropriate for the half-resolution and quarter-resolution situations, respectively.

The most efficient use is made of the channel when the samples of the filter output are taken at the Nyquist rate, i.e., at twice the highest frequency present in the filter output. Designating the sampling rate at full resolution as  $f_s$ , it is then clear that at half resolution and quarter resolution, the Nyquist sampling rate required is only  $f_s/2$  and  $f_s/4$ , respectively. Since the image is sampled in two spatial dimensions, this corresponds to a reduction in information rate over the channel by a factor of  $2^2 = 4$  or  $4^2 = 16$  for decimation factors of two and four, respectively. Since the recording medium is always exposed at the number of samples per inch corresponding to the highest system resolution, the decimated sample sequence must be interpolated up to this rate. Decimation and interpolation were discussed in detail in Section 4.1.3. The decimation ratio is designated by  $D$  in Figure 6.3-1 and the horizontal decimation/interpolation process is explicitly shown in the figure. The vertical decimation is accomplished by merely skipping the appropriate number of scan lines. The implementation of the vertical (cross-scan) interpolation process is more involved and is shown in more detail in Figure 6.3-2 for the case of interpolating by a factor of two with a FIR filter of length seven. Note that several of the tap weights ( $a_1, a_3, a_5$ ) are associated with zero-valued pixels in the figure. All of the tap weights are necessary since when reconstructing the following line, tap weights  $a_0, a_2, a_4$ , and  $a_6$  will then be associated with zero-valued pixels.



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Figure 6.3-2  
Detailed Description of  
Vertical Interpolation Process at the  
Recorder for an Interpolation Factor of Two

Notice that for this simple example that alternate lines are reconstructed using odd and even numbers of nonzero-valued tap weights into the summing junction of the interpolating filter. Since these nonzero-valued pixels at the interpolating filter input have the same statistics, i.e., mean and variance, regardless of line location, the effect of using a different number of effective tap weights on alternate lines is to modulate the filter gain from line to line. Since this same effect is present in the interpolation process in the orthogonal dimension, the result is that the reconstructed image will appear to have a grid-like structure superimposed on it. For filters having a large value of  $N$ , the difference between  $(\frac{N+1}{2})$  and  $(\frac{N+1}{2} - 1)$  is small and the corresponding modulation of filter gain is negligible, i.e., the grid-like pattern is not visible. This effect is much more pronounced when interpolating by a factor of four and is evident in some of the  $N=9$  filtered images discussed in paragraph 4.0. This effect is ameliorated by using a zero-mean pixel value representation system. See Appendix J for a detailed explanation of this effect.

The memory requirement for the cross-scan interpolating scheme of Figure 6.3-2 may be computed as follows. The number of pixels per line to be stored is  $(384 \text{ pixels/in})(9 \text{ in}) = 3456 \text{ pixels/line}$  and the maximum number of lines required to be stored is  $(N+1)/2$  as may be seen from the example of Figure 6.3-2. For the filter sizes used in the simulations  $N=9, 15$ , and  $31$ , this implies 17,280 words, 27,648 words, and 55,296 words of storage respectively, and at 4 bits/pixel this translates into a recorder memory requirement of 69.1 Kbits, 110.6 Kbits, or 2.2 Mbits for  $N=9, 15$ , or  $31$ , respectively. These memory requirements add substantially to the implementation complexity of this approach. The candidate system to be presented in the next section retains the REARCS capability of the system of Figure 6.3-2 while eliminating the massive memory requirement.



SEPARABLE TWO-DIMENSIONAL FILTERING, ONE-TAP  
DPCM, MTF CORRECTION, REARCS- SELECTABLE RESOLU-  
TION AND BWCR, MULTIPLE-ELEMENT RECORDING

In this section we present the logical extension of the candidate system presented in Section 6.3. The transmitting and receiving architectures of this multiple-element recording approach are shown in Figures 6.4-1 and 6.4-2, respectively. In both these figures only one set of tap weights is shown with each TAD, when in fact, three sets of tap weights are required for the three different resolutions. The alternate sets of tap weights were left out of the figures for clarity, but are required for actual implementation. In the transmitter architecture of Figure 6.4-1 the vertical portion of the separable filtering is performed for  $M$  scan lines simultaneously rather than for only one scan line. The number of tap weights per summing junction is designated by  $K$  and  $N$  is the length of the linear array sensor. The value of  $K$  in Figure 6.4-1 was chosen as  $K=5$  for clarity; in an operational system  $K$  must be larger, perhaps  $K > 15$ , in order to achieve acceptable results. The horizontal portion of the separable filtering is performed using  $M$  separate TAD's, each of length  $K$ ; the output of each of these FIR filters is then decimated by the horizontal decimation factor  $D_h$  prior to being stored in a P/S converter of length  $M$ . The contents of the P/S converter are clocked out sequentially, explicitly decimated by the vertical decimation factor  $D_v$ , A/D converted and encoded for transmission over the channel by a multiplexed DPCM encoder. Note that the

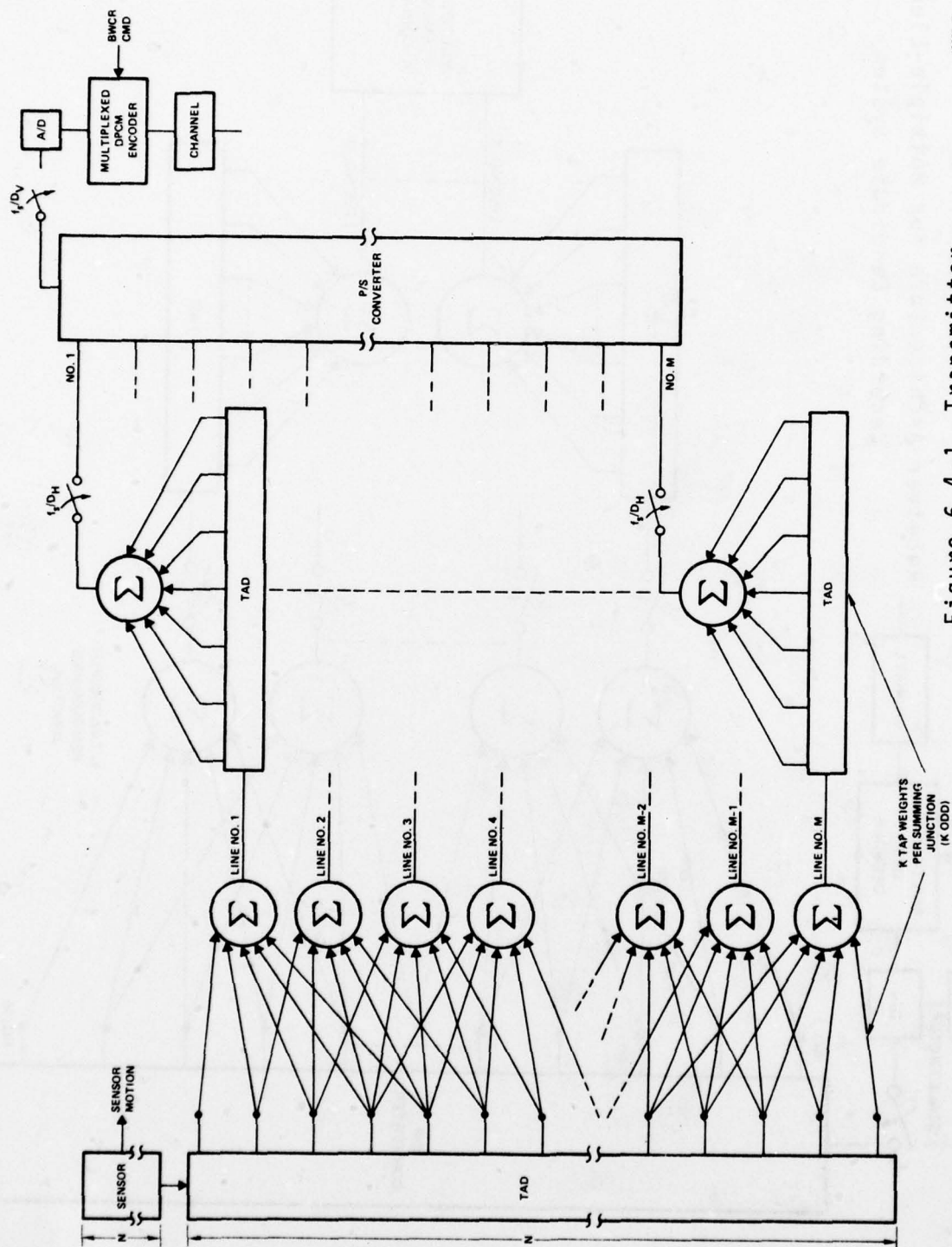
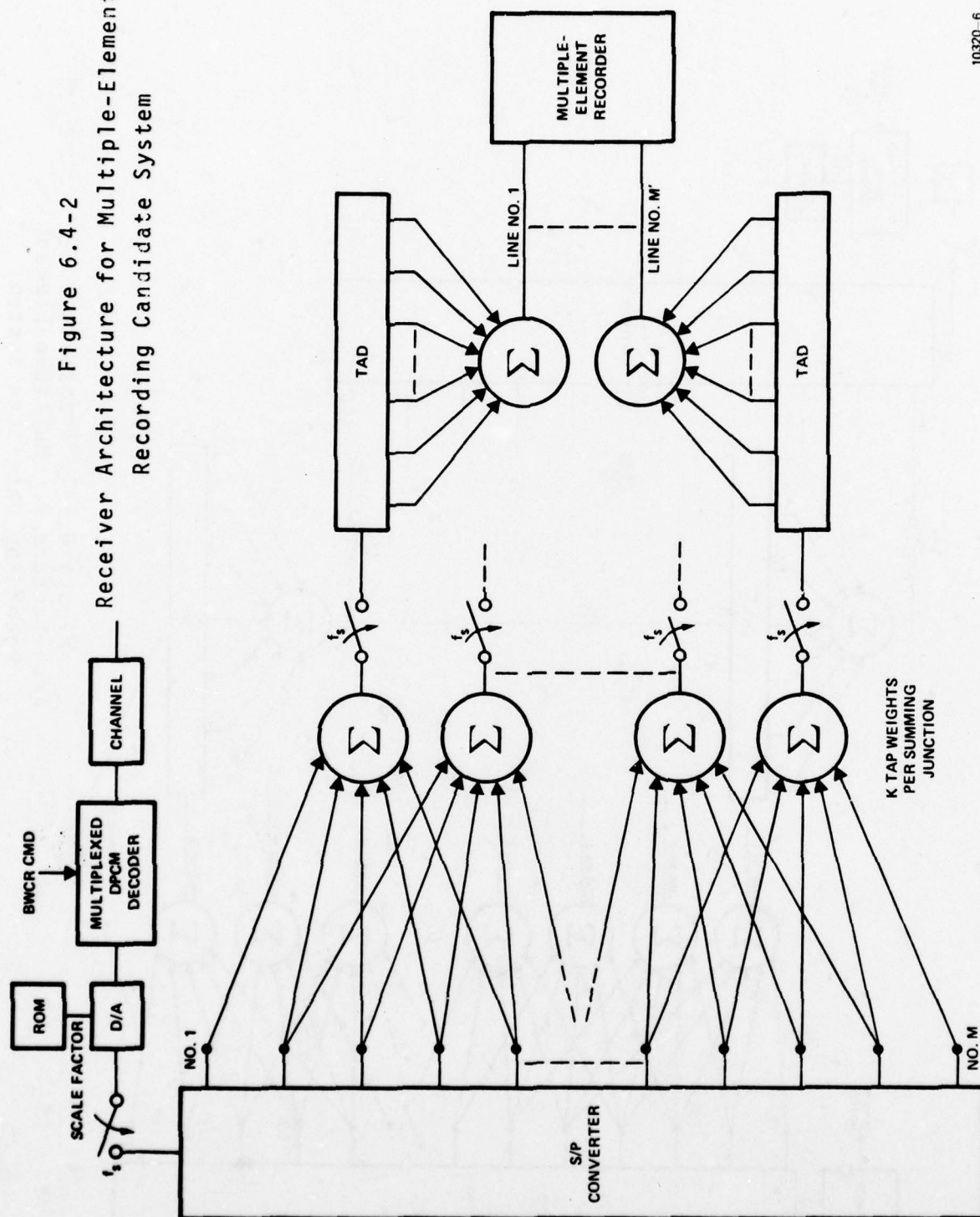


Figure 6.4-1. Transmitter Architecture for Multiple-Element Recording Candidate System

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**Figure 6.4-2**  
**Receiver Architecture for Multiple-Element**  
**Recording Candidate System**



vertical and horizontal decimation factors  $D_h$  and  $D_v$  are identical for the REARCS technique; the subscripts are used here only for added clarity. Also, note that  $K$ ,  $M$ , and  $N$  are related by

$$N = M + 2 \left( \frac{K+1}{2} - 1 \right) = M + K - 1 \quad (6-2)$$

At the receiver the sequence of operations is the reverse of those at the transmitter. This is shown in Figure 6.4-2, where the received digital data is decoded using a multiplexed DPCM decoder prior to being D/A converted. The output samples of the D/A converter are interleaved with zero-valued samples in such a way that the vertical sampling rate at the input to the S/P converter is again  $f_s$ . For  $D_v = 2$  or  $D_v = 4$  this implies inserting one or three zero-valued samples, respectively, between each pair of D/A output samples. This process is symbolically represented in Figure 6.4-1 by a switch at the output of the D/A closing at the vertical sampling rate  $f_s$ , even though the D/A output samples occur at the rate  $f_s/D_v$ . This symbolic notation will be used throughout the remainder of this report to represent the process of inserting zero-valued samples between nonzero-valued samples. Note that the vertical and horizontal sampling rates are identical and that both are designated by  $f_s$  throughout this report.

The contents of the S/P converter, including the zero-valued samples inserted at the output of the D/A process, are used in conjunction with the  $M'$  summing junctions to perform the vertical interpolation process. The output pixels of each of these  $M'$  vertical interpolation filters are then interspersed with the appropriate number of zero-valued samples prior to the horizontal interpolation process.



The horizontal interpolation process is performed by  $M'$  distinct FIR filters using TAD's, and the  $M'$  filter outputs then drive a multiple element recorder to complete the reconstruction of the image. The values of  $M$ ,  $M'$ , and  $K$  are related by

$$M = M' + K - 1. \quad (6-3)$$

Note that, unlike the candidate system of Section 6.3, the recording structure of this candidate system can take advantage of the fact that some pixels are zero-valued by deleting the tap weights that are always associated with zero-valued pixels. Recall that in the previous candidate system the same set of tap weights in effect, moved vertically across pixel values, and each tap weight therefore was not associated exclusively with either zero-valued or nonzero-valued pixels. For the present candidate system, this is not the case and approximately one-half or one quarter of the tap weights can be deleted from the vertical interpolating filters according to whether the interpolating factor is two or four.

The DPCM encoder and decoder must operate in a multiplexed fashion since they must process pixel values corresponding to  $M$  rows over a channel that is serial in nature. A block diagram of the multiplexed DPCM encoder and decoder is shown in Figure 6.4-3. In this figure the blocks labeled  $D_1$ ,  $D_2$ , ... represent the one-sample delays required in one-tap DPCM. For this multiplexed one-tap DPCM system each delay element would be realized as one word of a digital memory.

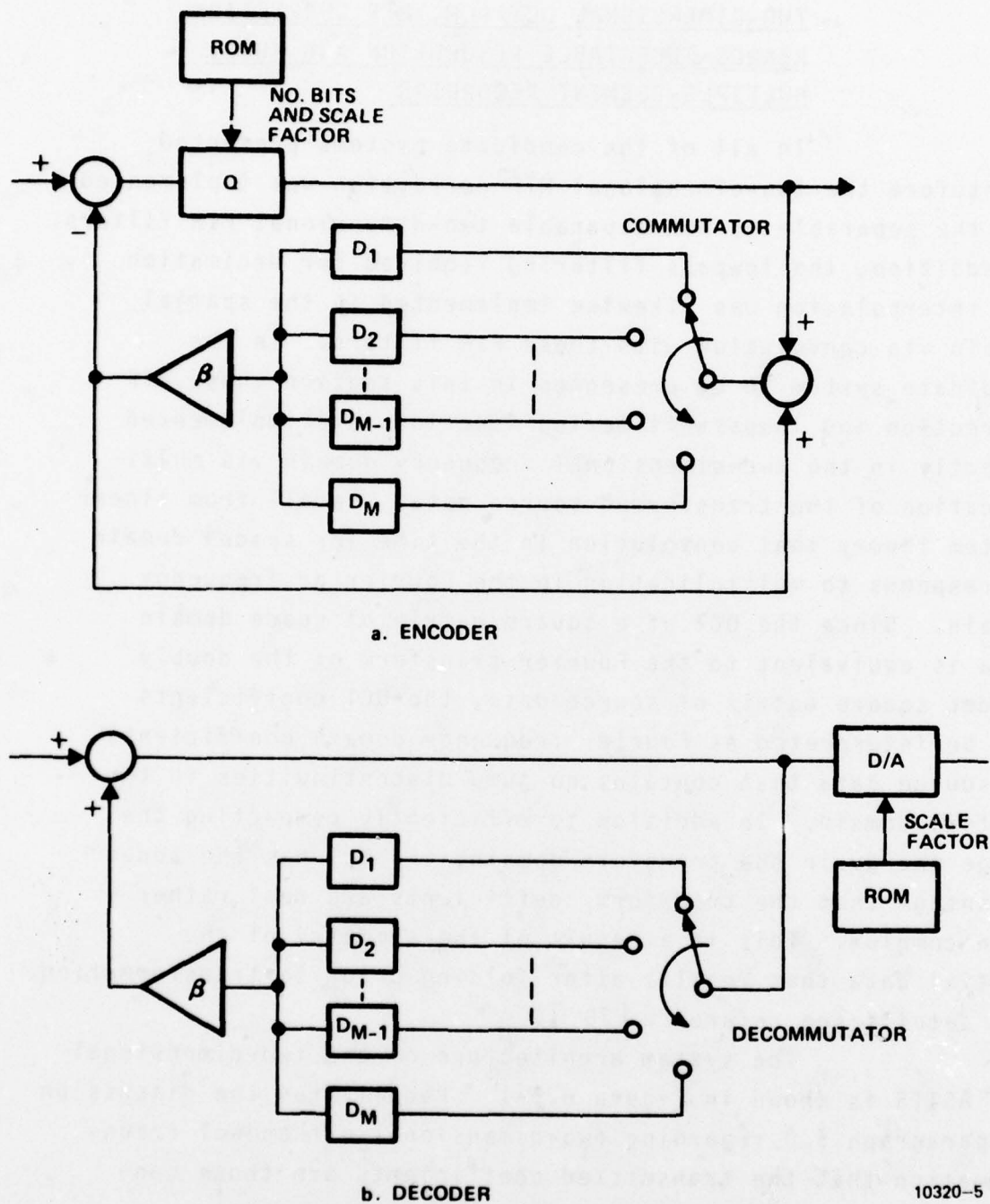
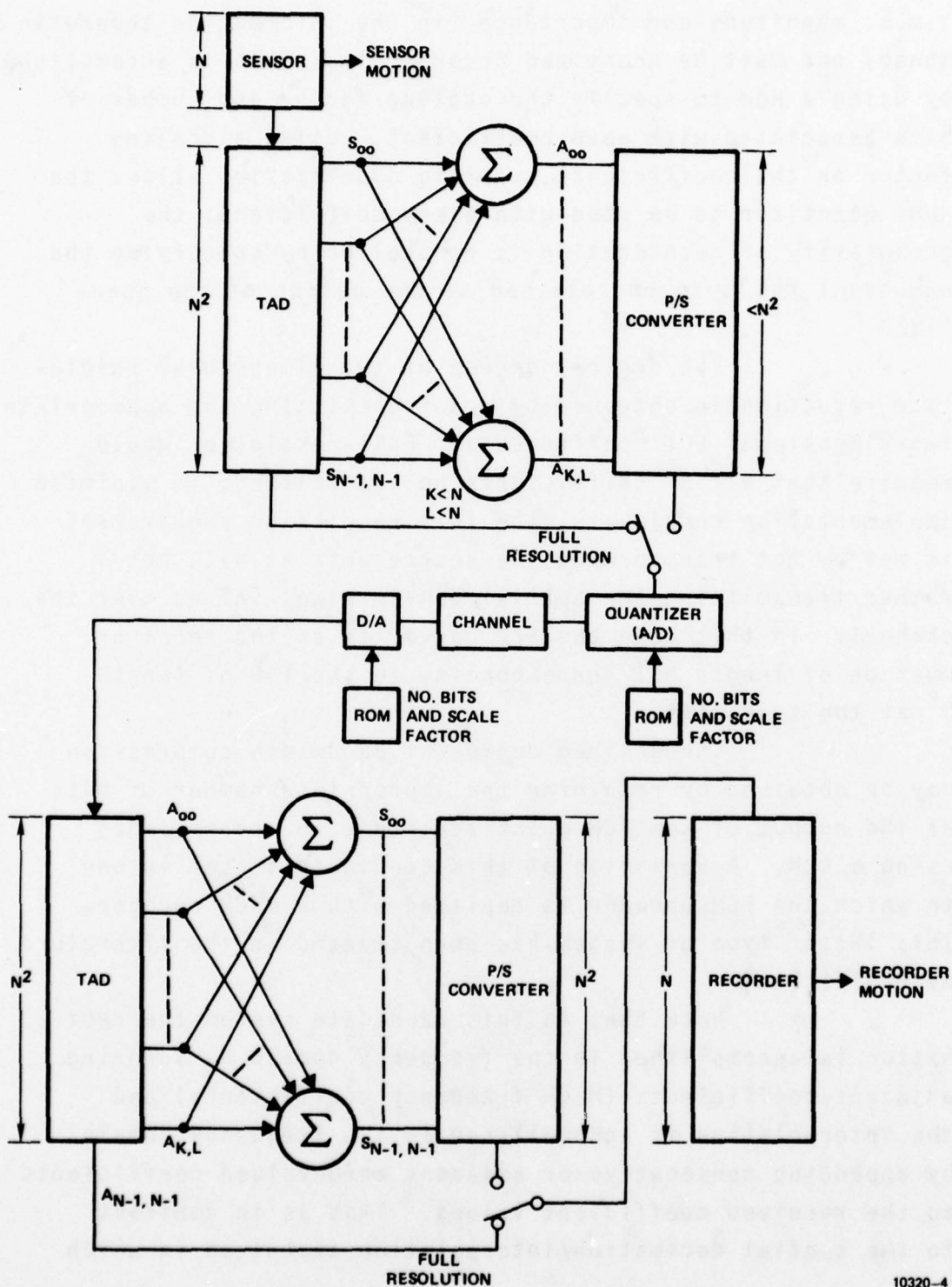


Figure 6.4-3  
Multiplexed DPCM Encoder/Decoder

TWO-DIMENSIONAL DCT/PCM, MTF CORRECTION,  
REARCS-SELECTABLE RESOLUTION AND BWCR,  
MULTIPLE-ELEMENT RECORDING

In all of the candidate systems presented heretofore the two-dimensional MTF correction was implemented via the separable or non-separable two-dimensional FIR filters. In addition, the lowpass filtering required for decimation and interpolation was likewise implemented in the spatial domain via convolution with these FIR filters. In the candidate system to be presented in this section these MTF correction and lowpass filtering functions are implemented directly in the two-dimensional frequency domain via multiplication of the transformed source data. Recall from linear system theory that convolution in the time (or space) domain corresponds to multiplication in the Fourier or frequency domain. Since the DCT of a square matrix of space domain data is equivalent to the Fourier transform of the doubly folded square matrix of source data, the DCT coefficients may be interpreted as Fourier frequency domain coefficients of source data that contains no jump discontinuities in the spatial domain. In addition to efficiently compacting the image energy in the transform domain, the DCT has the added advantage that the transform coefficients are real rather than complex. This is a result of the symmetry of the spatial data that results after folding prior to transformation. For details see reference [30 ].

The system architecture of the two-dimensional DCT ASITS is shown in Figure 6.5-1. Recall from the discussion of paragraph 5.0 regarding two-dimensional orthogonal transformation that the transmitted coefficients are those contained within certain zones defined by their manhattan distance from the origin. These coefficients differ in



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Figure 6.5-1. Two-Dimensional DCT/PCM With Multielement Recorder



r.m.s. magnitude and importance (in the information theoretic sense) and must be quantized accordingly. This is accomplished by using a ROM to specify the scaling factor and number of bits associated with each coefficient. Using a scaling factor on the coefficients prior to quantization allows the same quantizer to be used with every coefficient; the granularity of quantization is controlled by specifying the number of MSB's to be retained at the output of the quantizer.

The desired degree of two-dimensional resolution reduction is obtained by not transmitting the appropriate two-dimensional DCT coefficients. Full resolution would require that all  $N^2$  coefficients be transmitted; to minimize implementation complexity, the full resolution requirement is met by not transforming the source data at all, but rather transmitting the spatial domain pixel values over the channel. In this case the S/P converter at the receiver must be of length  $N^2$ , corresponding to the TAD of length  $N^2$  at the transmitter.

The desired degree of bandwidth compression may be obtained by retaining the appropriate number of bits at the output of the PCM quantizer; this is accomplished using a ROM. A variation of this candidate system is one in which the PCM encoder is replaced with a DPCM encoder. This latter type of system has been treated in the literature by Habibi [16].

Note that in this candidate system the decimation is accomplished in the frequency domain by ignoring adjacent coefficients (high frequency coefficients) and the interpolation is accomplished in the frequency domain by appending consecutive or adjacent zero-valued coefficients to the received coefficient values. This is in contrast to the spatial decimation/interpolation technique in which

the zero-valued pixels were alternated with nonzero-valued pixels.

Before presenting an alternative realization of the two-dimensional DCT/PCM technique, refer once more to Figure 6.5-1. If the TAD device shown there is assumed to be the Reticon TAD-32, the size of the transform is limited to  $5 \times 5$ . By cascading two such Reticon TAD-32 devices for an overall length of 64, one can implement the commonly used  $8 \times 8$  DCT in order to obtain better BWC performance and higher flexibility regarding BWCR and image resolution. Whether one or two TAD-32's is used, it is necessary to include buffer amplifiers (not shown) at each stage of the TAD in order to properly drive the input resistors of the operational amplifiers that perform the weighting and summing operations indicated in Figure 6.5-1. An alternative to using one (or two) TAD-32's and associated buffer amplifiers at each tap to drive the various weighting-and-summing operational amplifiers is to utilize a number of mask-programmable BBD's such as the Reticon R5602, as shown in Figure 6.5-2. Each of the R5602's is mask programmed with a different two-dimensional DCT basis function, with each adjacent column of the basis function occupying consecutive blocks of the  $N^2$ -length mask. Choosing  $N=8$  gives good BWC performance and also utilizes all 64 stages of the R5602 BBD. Since the DCT is its own inverse within a scale factor, the same DCT architecture shown in Figure 6.5-2 for the transmitter could also be utilized at the receiver.

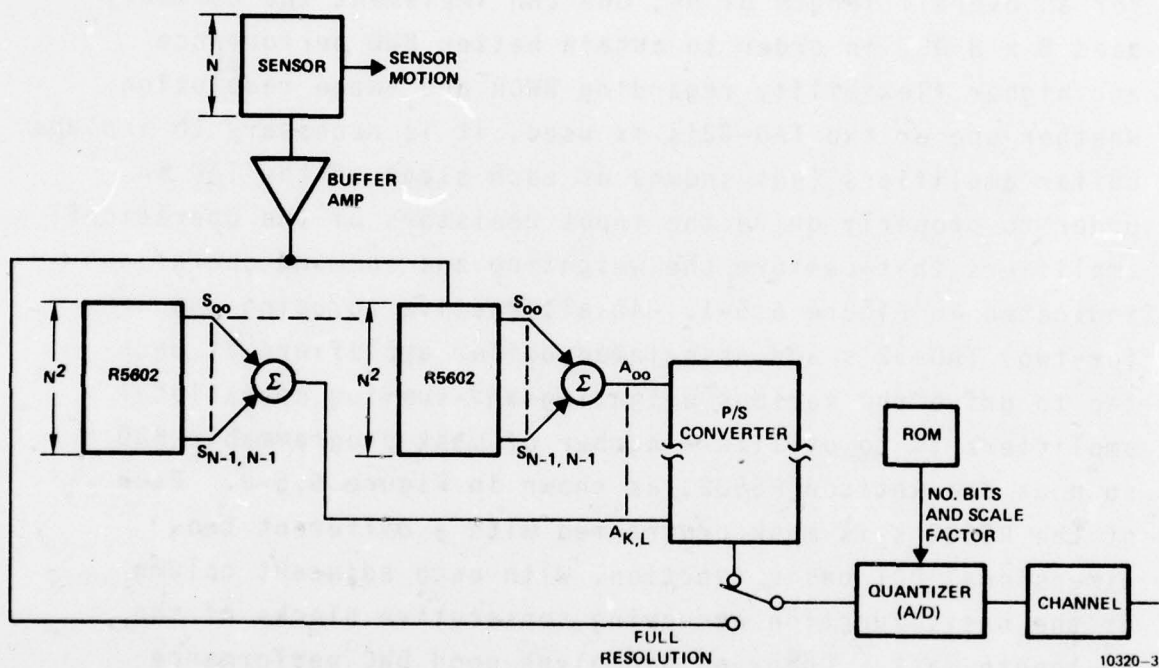


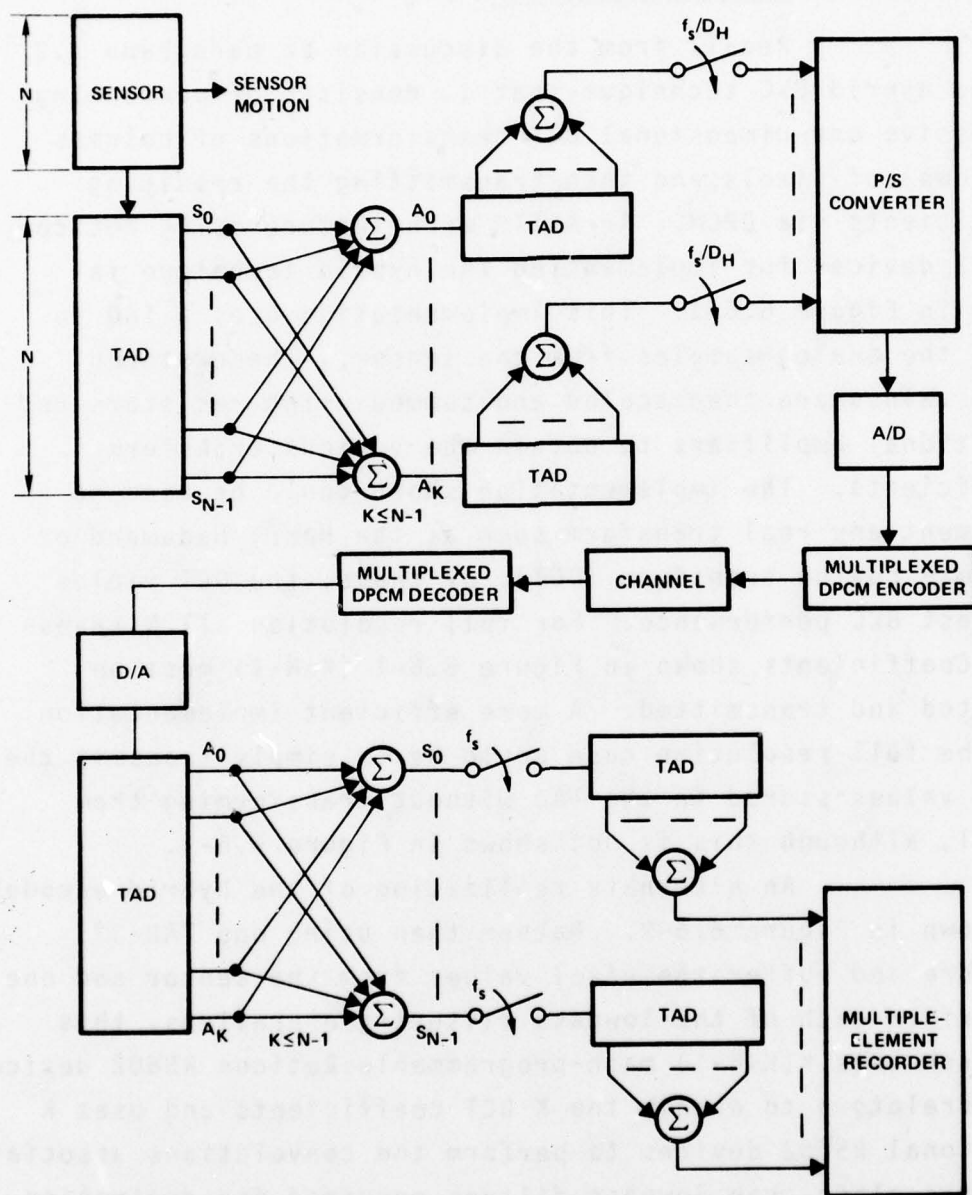
Figure 6.5-2  
 Alternate Architecture for Realization of  
 Two-Dimensional DCT Using R5602's for  $N \leq 8$

HYBRID DCT/DPCM, MTF CORRECTION, REARCS-  
SELECTABLE RESOLUTION AND BWCR, MULTIPLE-  
ELEMENT RECORDING

Recall from the discussion in paragraph 5.0 of the hybrid BWC technique that it consists of performing successive one-dimensional DCT transformations of columns (or rows) of pixels and then transmitting the resulting coefficients via DPCM. An ASITS architecture using Reticon TAD-32 devices for implementing the hybrid technique is shown in Figure 6.6-1. This implementation uses a TAD to store the analog samples from the sensor. These stored pixel values are then scaled and summed using resistors and operational amplifiers to obtain the various transform coefficients. The implementation shown could be used to implement any real transform such as the Haar, Hadamard or discrete cosine transform (DCT); of these, the DCT yields the best BWC performance. For full resolution all  $N$  transform coefficients shown in Figure 6.6-1 ( $K=N-1$ ) must be computed and transmitted. A more efficient implementation for the full resolution case would be to simply transmit the pixel values stored in the TAD without transforming them at all, although this is not shown in Figure 6.6-1.

An alternate realization of the hybrid encoder is shown in Figure 6.6-2. Rather than using one TAD-32 to store and buffer the pixel values from the sensor and one to perform each of the lowpass filtering operations, this approach uses  $K$  ( $K < N-1$ ) mask-programmable Reticon R5602 devices as correlators to obtain the  $K$  DCT coefficients and uses  $K$  additional R5602 devices to perform the convolutions associated with the along-scan lowpass filters required for decimation. This alternate realization using R5602 devices has lower





10320-2

Figure 6.6-1  
ASITS Architecture for Hybrid BWC Technique

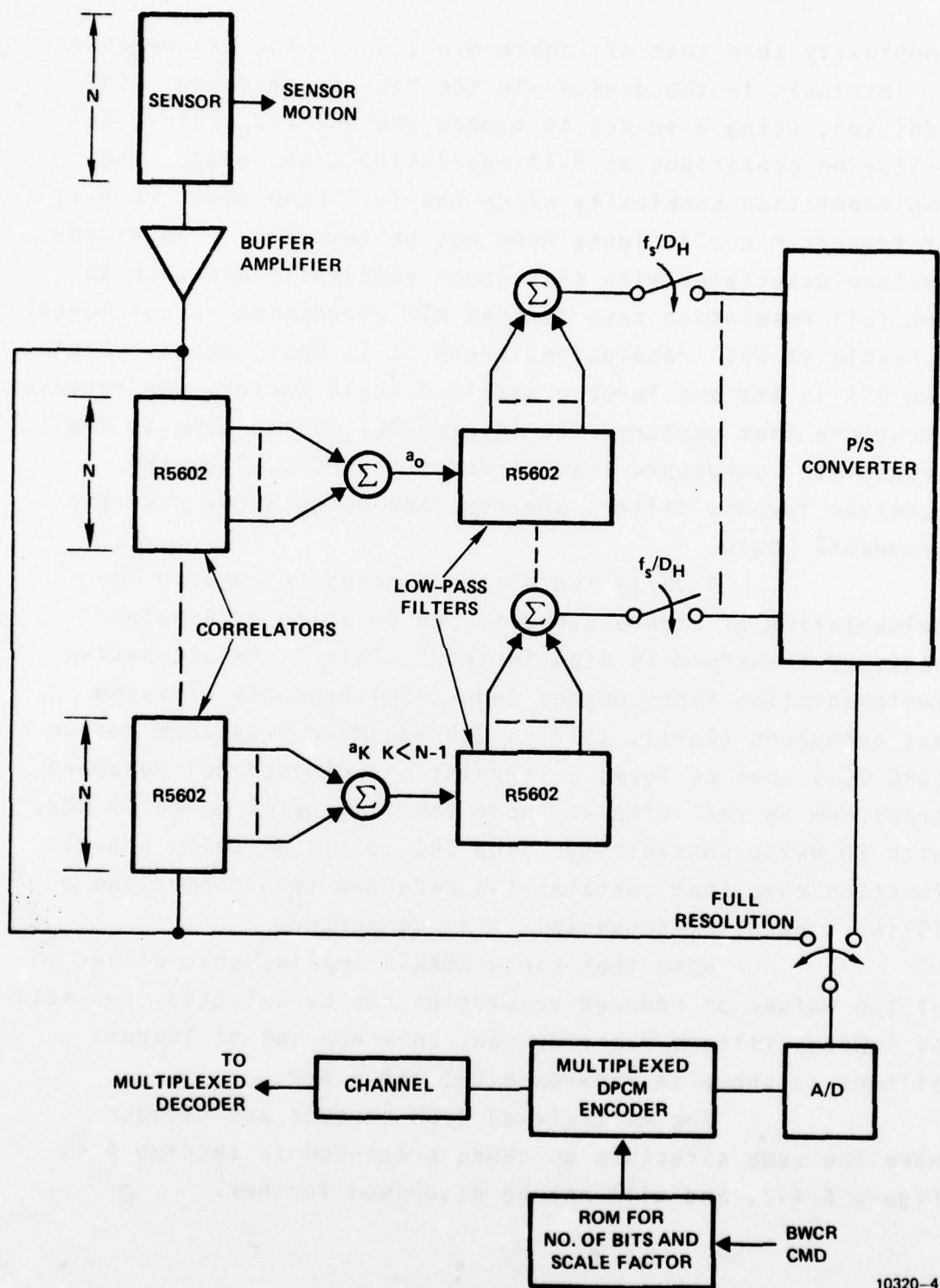


Figure 6.6-2 Alternate Realization of Hybrid Encoder  
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complexity than that of Figure 6.6-1 since the tap weighting is intrinsic to the device via the mask programming. In addition, using a switch to bypass the DCT and transversal filtering operations at full resolution also reduces the implementation complexity since the full complement ( $K=N-1$ ) of transform coefficients need not be computed. The disadvantage associated with this lower complexity approach to the full resolution case is that MTF correction is not longer possible at full resolution, where it is most needed. Since the DCT is its own inverse within a scale factor, the receiver structure that performs the inverse DCT is the same as the transmitter structure that performs the DCT; also, the receiver lowpass filters are realized using R5602 mask programmable CCD's.

A variation of the N-point DCT hybrid implementation of Figure 6.6-2 is one in which a 16-point Hadamard transform is used instead. This is an attractive implementation since Hughes Industrial Products Division has announced (April, 1977) a CCD Hadamard Transform device (CRC 101) that performs a 16-point one-dimensional Hadamard transform in real time at input sampling rates up to 10 MHz, with 20 MHz a possibility. The CRC 101 is actually a dual function chip that contains the Hadamard transformer and a 16-input parallel-to-serial (P/S) converter.

Note that since REARCS implies that either of two values of reduced resolution can be selected, two sets of lowpass filters are required; only one set of lowpass filters is shown in Figures 6.6-1 and 6.6-2.

The multiplexed DPCM encoder and decoder have the same structure as those presented in Section 6.4, Figure 6.4-3, and will not be discussed further.

As was pointed out in Section 6.5 for two-dimensional transform techniques, the resolution can be selected or controlled in the frequency domain as well as in the spatial domain. In the hybrid BWC approach shown in Figures 6.6-1 and 6.6-2 the spatial resolution in the vertical spatial dimension is controlled by the number of transform coefficients  $a_j$  that are transmitted over the channel; discarding more of the high order coefficients in the frequency domain decreases the spatial resolution within the image after inverse transformation. Unlike the two-dimensional transform technique, however, the horizontal spatial resolution is controlled by FIR transversal filters whose inputs are the along-scan spatial sequence of transform coefficients and whose outputs are decimated by the appropriate decimation factor. It is clear that in this hybrid ASITS the vertical spatial resolution is controlled in the frequency domain by eliminating transform coefficients while the horizontal spatial resolution is controlled in the spatial domain via lowpass filtering using transversal filters. This approach is appropriate in view of the mixed-domain nature of the two-dimensional number sequence that is transmitted over the channel.

#### 6.7

#### ALL-DIGITAL IMPLEMENTATION OF HYBRID DCT/DPCM BWC TECHNIQUE

In all of the candidate systems presented heretofore the implementations have been largely sampled-data analog in nature, with very little digital processing. This partitioning of the processing between analog and digital implementations is generally advantageous in that the highly computation-intensive two-dimensional filtering tasks at the transmitter and receiver can be accomplished using sophisticated, commercially available CCD's that can be either externally programmed (TAD-32) or mask-programmed



(R5602). In order to help evaluate the relative merits of these predominantly analog (sampled-data) ASITS implementations that have been presented we present several alternative all-digital implementations of the basic hybrid DCT/DPCM BWC technique without the REARCS capability for specifying two-dimensional resolution. Recall from the earlier discussion of Section 6.6 that for the hybrid technique the vertical resolution is determined by the number of DCT coefficients that are retained for transmission over the channel, while the horizontal resolution is reduced by lowpass filtering each of the retained DCT coefficient sequences. Thus, the all-digital hybrid algorithm implementation of this section has implicit control capability for vertical resolution but no control capability for horizontal resolution.

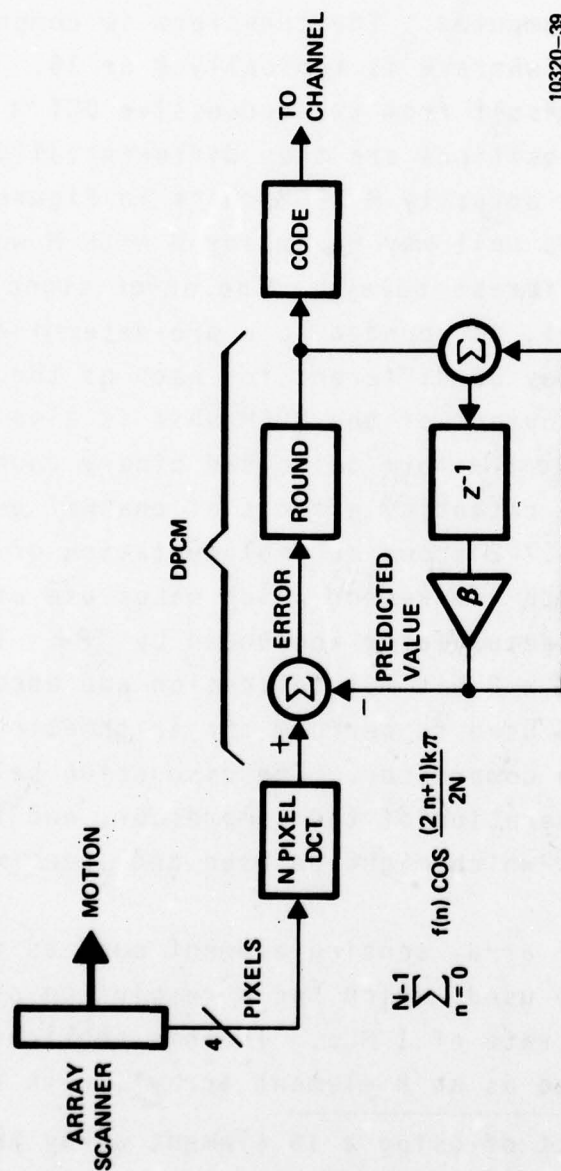
Several different digital implementations of a hybrid DCT/DPCM bandwidth compression system are discussed in this section. The options presented are intended to provide a broad overview of the range of digital implementations which are possible.

Figure 6.7-1 shows the functional processing steps involved in the hybrid DCT/DPCM bandwidth compression approach. The DCT itself is defined to be

$$A_k = \sum_{n=0}^{N-1} f(n) \cos \frac{(2n+1)K\pi}{2n}, \quad (6-4)$$

for

$$k = 0, 1, 2, \dots, N-1, \quad (6-5)$$



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Figure 6.7-1  
Functional Block Diagram of the Hybrid BWC Approach

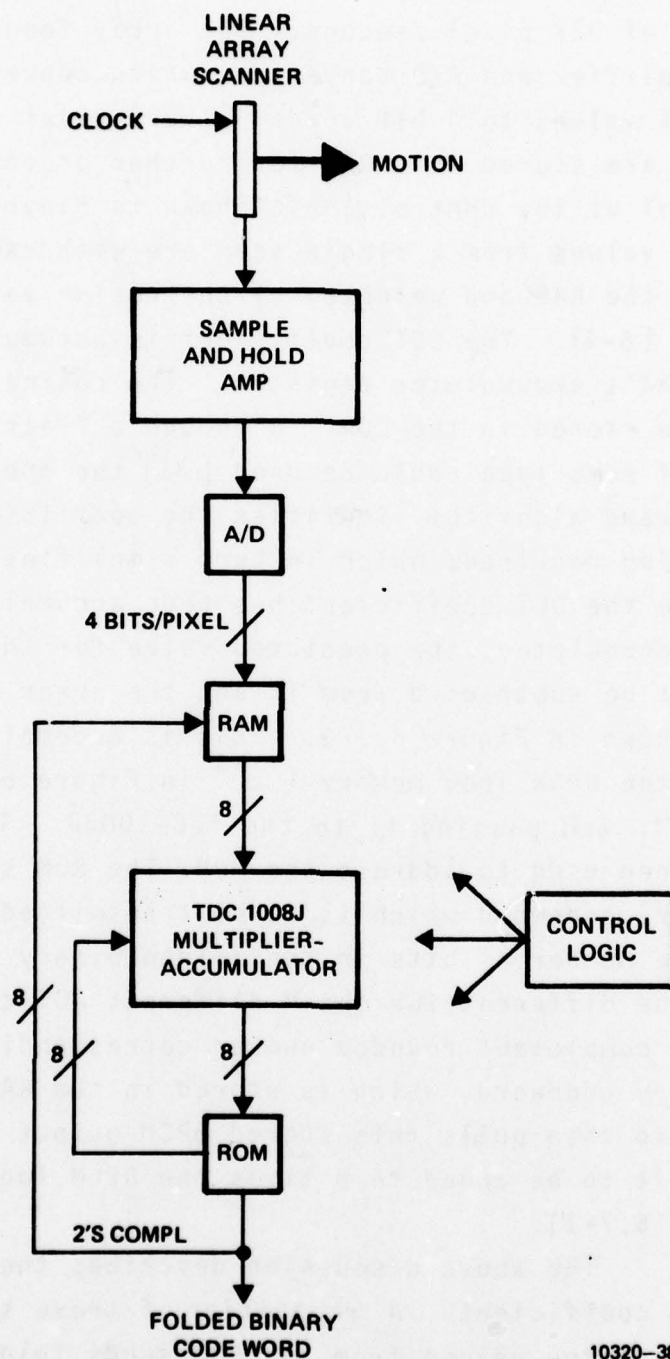
where  $f(n)$  is the pixel value, and  $N$  is the number of points in the transform. The transform is defined for  $N$  values of  $k$ , but typically only the lower indexed coefficients need be computed. In the following it is assumed that  $M < N$  coefficients are actually computed. The transform is computed on a block of  $N$  pixels, where  $N$  is typically 8 or 16. The  $2M$  coefficients which result from two successive DCT's computed at successive scanner positions are then differentially encoded. Hence there are actually  $M$  DPCM units in Figure 6.7-1, although one time-shared unit may be employed with  $M$  words of memory for the  $M$  different delays. The error signal, which is already digital, is rounded to a pre-determined number of bits, which may be different for each of the  $M$  DCT coefficients. The output of the DPCM unit is also typically re-coded from 2's complement form to folded binary codeword form [30] to reduce the potential effects of channel errors.

Figure 6.7-2 shows an implementation of the hybrid DCT/DPCM bandwidth compressor which makes use of the single chip multiplier/accumulator announced by TRW. The TDC 1008J can perform  $8 \times 8$  bit multiplication and accumulation in 70 nsec, and is used to perform the arithmetic required in the bandwidth compressor. The discussion below explains the overall operation of the compressor, and identifies typical components which might be used and speed/memory requirements.

A linear array sensing element such as the Reticon RL-64P could be used, which has a resolution of 500 lpi and a maximum data rate of 1 MHz. In this application the RL-64P would be used as an 8 element array\*, with a pixel

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\* In general, the effect of using a 16 element array and transform is to double the memory requirements given subsequently. The speed of the devices used in most of the approaches discussed below is quite adequate to handle the higher throughput required by the 16 point DCT.



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Figure 6.7-2  
 Hardware Implementation Using TDC 100BJ  
 Multiplier/Accumulator  
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output rate of 32K pixels/second. The array feeds a sample-and-hold amplifier and A/D converter, which converts the analog pixel values to 4 bit words. The 8 pixel values from a scan are stored in a RAM for further processing. Under control of the control logic shown in Figure 6.7-2, the 8 pixel values from a single scan are withdrawn sequentially from the RAM and weighted by the cosine values as in equation (6-4). The DCT coefficient is accumulated in the TDC-1008J's accumulator register. The cosine values required are stored in the ROM. Although a "fast" DCT algorithm of some type could be used [31] the speed of the straightforward algorithm simplifies the coefficient and pixel indexing required, which in turn simplifies the control logic. Once the DCT coefficient has been accumulated in the TDC 1008J accumulator, the predicted value for the coefficient must be subtracted from it and the error rounded and coded, as shown in Figure 6.7-1. This is accomplished by retrieving the DPCM loop memory ( $z^{-1}$  in Figure 6.7-1) from the RAM, and passing it to the TDC-1008J. The accumulator output is then used to address the ROM. The ROM supplies the folded binary codeword which is to be transmitted over the channel, the number of bits in the folded binary codeword (which may be different for the M different DCT coefficients), and the 2's complement rounded number corresponding to the folded binary codeword, which is stored in the RAM. The control logic then pulls this stored DPCM output from RAM and causes it to be added to  $\beta$  times the DPCM loop memory (see Figure 6.7-1).

The above discussion describes the processing for one DCT coefficient. A repetition of these steps with appropriate cosine values from the ROM sends folded binary codewords over the channel for the remaining (M-1) DCT coefficients.

The TDC-1008J must compute M DCT coefficients per 8 new pixels accepted from the scanner. At 32K pixels/second, this is a rate of M DCT coefficients per 250 microseconds. If it were necessary to compute all 8 coefficients (M=8), this is 31 microseconds per coefficient worst-case. Since each coefficient requires 8 multiply-adds using a "non-fast" algorithm, the required worst-case multiply-add time is  $31/8=3.9$  microseconds, which is well within the capabilities of the TDC-1008J (70 nsec).

The RAM must be able to hold the 8 pixel values which are currently being processed, plus the 8 new pixel values which are being placed in memory while the current pixels are processed. It must also hold the M DPCM loop memories, which is at most 8 words for M=8. Allowing for a small scratch-pad area, a RAM of 32 words x 8 bits should suffice.

The ROM must hold the 8M cosine values (to make the control logic as simple as possible no symmetry is used), which for M=8 is a maximum of 64 cosine values. The ROM must also hold the folded binary codewords, the 2's complement rounded values, and the number of bits in the folded binary codeword for each DCT coefficient. If we assume that M=6, and that 16 bits are required to retain the above information, then the number of 8 bit words of storage required is

$$2^8 \times 6 \times 2 + 64 = 3136 \quad (6-6)$$

The  $2^8$  factor results from the 8 bit accumulator output used to address the ROM, the factor of 6 results from the 6 different DCT coefficients, the factor of 2 results from the 2 words (16 bits) of storage required for each access, and

the 64 is the number of cosine values which must be stored. The required storage can be realized in a single chip such as Texas Instruments' TMS 4732, which has a 4096 word x 8 bit organization.

Careful consideration of the implementation described above reveals that the hard-wired control logic must be fairly complex in order to properly sequence the compressor operation, provide ROM memory addresses, etc. An alternative approach is to use a microprocessor for control, such as the AMD 2901. Figure 6.7-3 shows a block diagram of a micro-processor-based bandwidth compressor. Due to the programmability of the micro-processor, the processor shown in Figure 6.7-3 could also be used to perform other image processing algorithms. The architecture is general purpose with special features added to enhance its signal processing capability. The processor is built around a central array of four 4-bit AM2901A processing elements. Emanating from the central array are two unidirectional data busses, each 16 bits wide.

Between these two busses are positioned:

- Data Memory (autoincrementing)
- High Speed Multiplier (16 x 16 in 200 ns)
- Saturation Circuitry
- Sign Extension Circuitry
- Bit Reversal Circuitry
- Bidirectional I/O Buss

The microprogram sequencer used to control fetches from a separate instruction memory is a single chip AMD 2901 device that provides a variety of conditional branching, looping and subroutine calls along with an internal loop counter. One level of pipelining is used in the 56-bit wide instruction stream and one level of program controlled interrupt is provided.

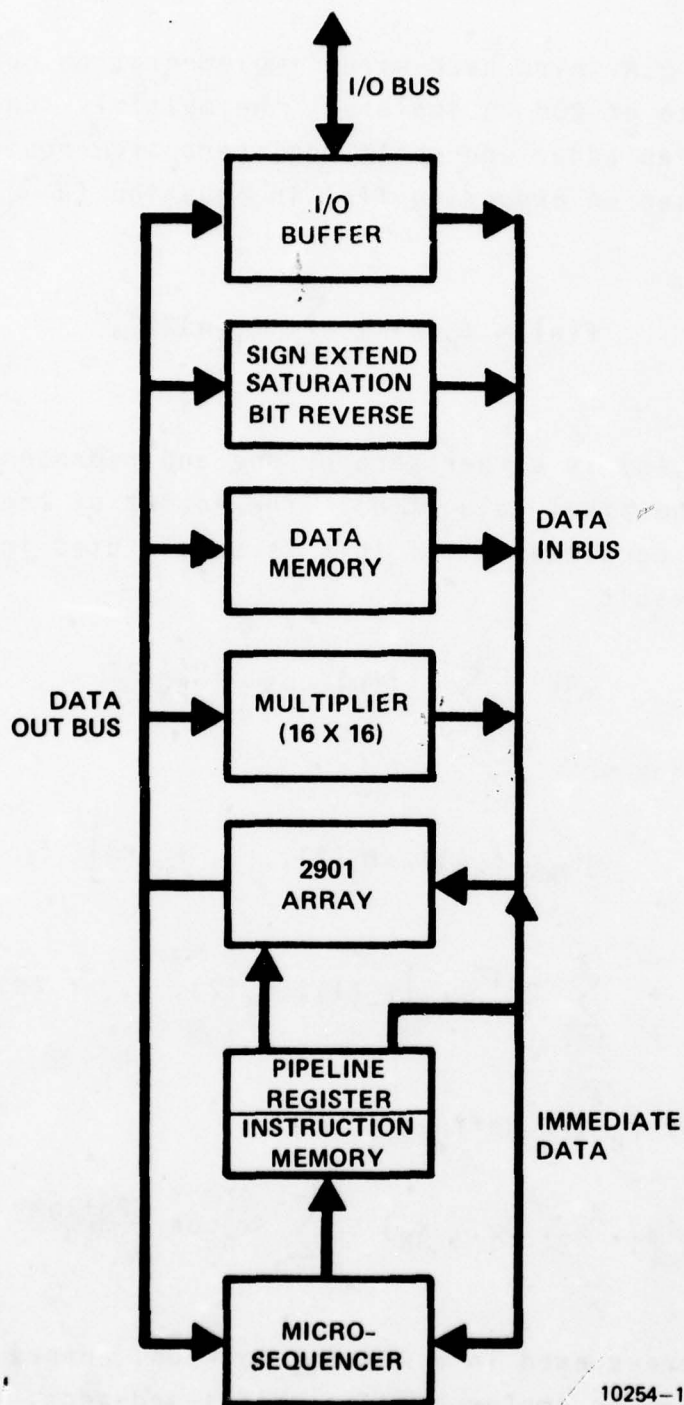


Figure 6.7-3.  
Block Diagram of a Micro-Processor-Based  
Bandwidth Compressor  
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A third hard-wired implementation option involves the use of ROM to implement the multiply function [32] so that only an adder and shift registers are required. The method is based on expanding  $f(n)$  in equation (6-4) in the form

$$f(n) = f_0(n) + \sum_{i=1}^3 f_i(n)2^{-i}, \quad (6-7)$$

where each  $f_i(n)$  is either zero or one and represents the  $i^{\text{th}}$  bit in the pixel value  $f(n)$ . The format of the pixel value is 2's complement. If this is substituted in equation (6-4), the result

$$\begin{aligned} A_k &= \sum_{n=0}^{N-1} f(n) \cos \frac{(2n+1)k\pi}{2N} = \\ &\phi_k [f_0(1), f_0(2), \dots, f_0(N)] \\ &+ \sum_{i=1}^3 2^{-i} \phi_k [f_i(1), f_i(2), \dots, f_i(N)] \end{aligned} \quad (6-8)$$

The functions  $\phi_k$  are defined to be

$$\phi_k(x_1, x_2, \dots, x_N) \equiv \sum_{n=0}^{N-1} x_n \cos \frac{(2n+1)k\pi}{2N} \quad (6-9)$$

and can be pre-stored in a ROM for look-up. Hence all operations can be implemented as shifts-and-adds, as is

apparent from (6-8). One of the disadvantages of this approach is the need to have the  $i^{\text{th}}$  bit of all  $N$  pixels available in parallel, which requires the use of tapped shift registers. The other approaches lend themselves more naturally to use word-organized memories which currently have a higher level of integration.

A fourth approach would employ ROM for table look-up of multiplications, thereby eliminating the need for a hardware multiplier such as the TDC-1008J. The ROM must be present in any event to provide the table look-up for the folded binary codewords, etc. If the pixel values and the DCT coefficient values are each 4 bits long, then  $2^8=256$  words of ROM memory are sufficient to provide a table of all possible multipliers and multiplicands and their resulting products. When added to the 3136 words of ROM required for the folded binary codewords, cosine values, etc., the total ROM required is 3392 words. This would fit within a single 4096 x 8 bit chip, such as Texas Instruments' TMS 4732. Note, however, that an adder, an accumulator, and various registers are required to utilize the ROM in the processing required for bandwidth compression. Hence the parts count may not be significantly less than that for some of the other approaches.

Considering the options presented above from an overall point-of-view, it would appear that the first two options are the best choices. The microprocessor-based option has the added feature of programmability, if the application requires it. The microprocessor approach may even have a lower overall parts count than the hardwired approach, depending primarily on the comparison which exists between the hard-wired control logic parts count and the software-based microprocessor parts count.

The Micro-Adaptive Picture Sequencing BWC technique was invented in 1977, and the relative effectiveness of this algorithm is the subject of current research. An array scanning implementation of the MAPS technique is presented here for the sake of completeness, although it does not exhibit many of the desirable features of an ASITS such as explicit control of resolution reduction or MTF control.

Figure 6.8 shows an array scanning implementation of the MAPS algorithm. Note that the implementation is all-digital with the A/D and D/A interfaces located directly at the array sensor output and array recorder input, respectively. The only advantage, other than higher image throughput rate, of this array scanning implementation over a conventional single-point scanning implementation is that the quantity of required data storage memory is much smaller. A simple example is illustrative. Consider a MAPS system for which the maximum data block size is constrained to be 8 pixels by 8 pixels. For the conventional single-point scanning approach this would require

$$384 \text{ pixels/in} \times 9 \text{ in} \times 8 \text{ lines} \times 4 \text{ bits/pixel} = 110.6\text{K bits} \quad (6-10)$$

of storage both at the transmitter and at the receiver. For an array scanning approach, on the other hand, the required amount of data memory is only

$$8 \times 8 \times 4 = 256 \text{ bits} \quad (6-11)$$

which is smaller by a factor of 432. Aside from the difference in required data storage the remainder of the digital implementation is the same regardless of whether single-point

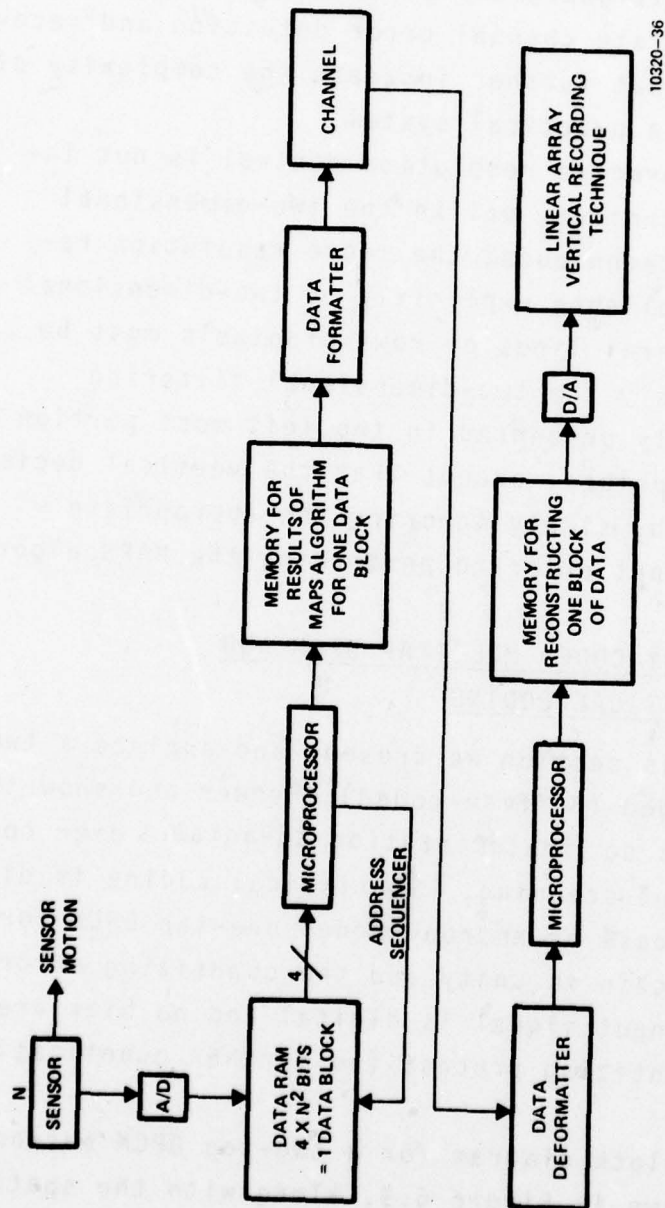


Figure 6.8  
All-Digital Architecture for Implementing MAPS Algorithm



scanning or array scanning is used.

One disadvantage of using the MAPS technique is that a channel buffer is required on both ends of the channel due to the nonconstant data rate of encoded output data; in addition to the additional memory associated with the channel buffers, elaborate channel error detection and recovery schemes are required that further increase the complexity of the implementation of a practical system.

Since precise resolution control is not inherent in the MAPS technique, unlike the two-dimensional transform and hybrid techniques, the image resolution reduction must be accomplished explicitly by two-dimensional filtering. Since several lines or rows of pixels must be processed simultaneously the two-dimensional filtering architecture previously presented in the left most portion of Figure 6.4-1 is appropriate, except that the vertical decimation is accomplished by simply ignoring the appropriate horizontal filter output prior to performing the MAPS algorithm.

## 6.9 ENTROPY-CODED MULTITAP DPCM AND STATISTICAL CODING

In this section we present and analyze a two-tap DPCM, entropy-coded (Huffman-coded) encoder and show that array scanning offers no implementation advantages over conventional single-point scanning. Statistical coding is discussed as a special case of entropy coded one-tap DPCM for which the predictor gain is unity and the quantizing error is zero, i.e., the input signal is digital and no bits are discarded in the quantizing process (no further quantization occurs).

The block diagram for a two-tap DPCM entropy coded encoder is shown in Figure 6.9, along with the spatial

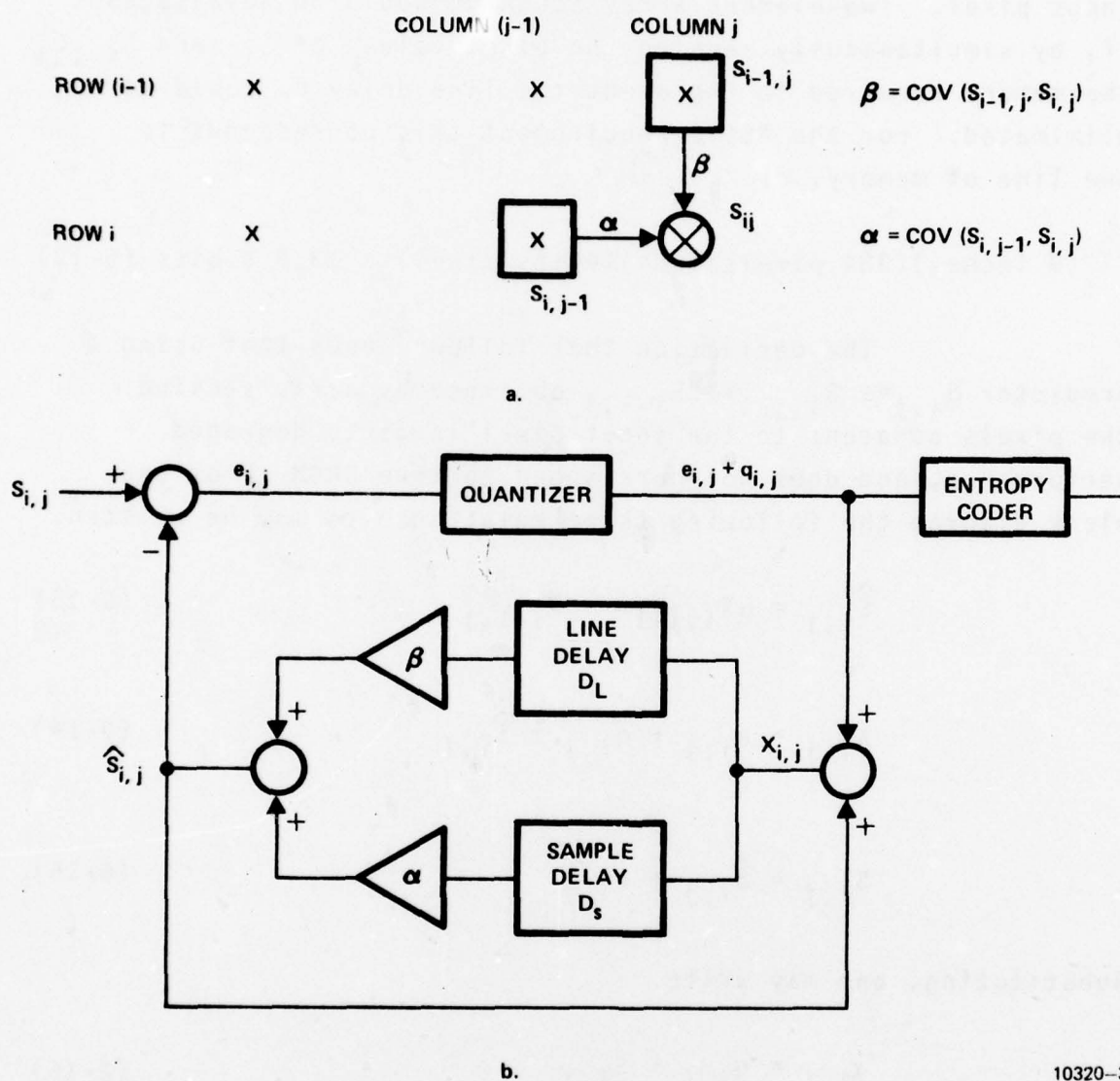


Figure 6.9  
Two-Tap Entropy-Coded DPCM

relationships among the input pixel or sample  $S_{i,j}$  and the two samples,  $S_{i-1,j}$  and  $S_{i,j-1}$ , used to predict the value of the input pixel. Two-element array scanning would be advantageous if, by simultaneously sensing the pixel values of  $S_{i,j}$  and  $S_{i-1,j}$  the memory required to implement the line delay  $D_L$  could be eliminated. For the ASITS requirement this corresponds to one line of memory, or

$$(9 \text{ inches})(384 \text{ pixels/inch})(4 \text{ bits/pixel}) = 13.8 \text{ K bits} \quad (6-12)$$

The derivation that follows shows that using a predictor  $S_{i,j} = \alpha S_{i,j-1} + \beta S_{i-1,j}$  obtained by array sensing the pixels adjacent to the input pixel leads to degraded performance, and does not correspond to true DPCM. From the block diagram the following three relationships may be written.

$$\hat{S}_{i,j} = \alpha X_{i,j-1} + \beta X_{i-1,j} \quad (6-13)$$

$$X_{i,j} = e_{i,j} + q_{i,j} + \hat{S}_{i,j} \quad (6-14)$$

$$S_{i,j} = \hat{S}_{i,j} + e_{i,j} \quad (6-15)$$

Substituting, one may write

$$X_{i,j} = q_{i,j} + S_{i,j} \quad (6-16)$$

and then

$$\hat{S}_{i,j} = \alpha [q_{i,j-1} + S_{i,j-1}] + \beta [q_{i-1,j} + S_{i-1,j}] \quad (6-17)$$

$$\hat{S}_{i,j} = \alpha S_{i,j-1} + \beta S_{i-1,j} + \{\alpha q_{i,j-1} + \beta q_{i-1,j}\}$$

It is clear from the last expression that the "two-dimensional" quantizing noise terms within the brackets must be included in the estimate, in addition to the scaled values of the adjacent pixels, in order to obtain true two-tap DPCM operation. Recall that the proposed predictor based upon two-element array scanning was

$$S_{i,j} = \alpha S_{i,j-1} + \beta S_{i-1,j} \quad (6-18)$$

This predictor differs from that required for true DPCM operation by the quantity  $\{\alpha q_{i,j-1} + \beta q_{i-1,j}\}$  and the SNR performance is degraded accordingly. Since the SNR performance improvement associated with going from one-tap DPCM to two-tap DPCM is slight to begin with, on the order of 1.9 dB according to theoretical calculations by O'Neal [19], the performance degradation that results from using the array scanning predictor discussed above probably negates this marginal increase in SNR performance due to going to a two-tap DPCM system. For this reason, the array scanning approach is not well suited for multitap DPCM systems.

Statistical coding is a special case of entropy-coded DPCM; for the entropy-coded two-tap DPCM system of Figure 6.9, setting  $\beta=0$ ,  $\alpha=1$  and eliminating the quantizer yields a statistical coder. Since statistical coding is a one-dimensional process array scanning offers no implementation advantages over conventional single-point scanning. In addition, statistical coding requires the additional implementation complexity associated with channel buffers and channel error recovery procedures. The channel buffers are required to accommodate the varying data rate at the output of the Huffman coder. One popular channel error recovery strategy replaces a line or line segment containing



the effects of a channel error with the previously decoded vertically adjacent line or line segment; this requires one line of storage plus the associated error detection and control circuitry. In view of all of the above considerations, it is clear that statistical coding is not appropriate for array scanning systems.

## 7.0 CONCEPTUAL DESIGN OF AN ARRAY SCANNING IMAGE TRANSMISSION SYSTEM

In this section the basis for selecting the hybrid DCT/DPCM candidate ASITS system for detailed design is discussed. The validity of the hybrid mixed-domain filtering approach used to control vertical and horizontal spatial resolution is illustrated with a simple example using a four-element DCT and this is followed by a mathematical proof for the general case of transforms of length  $N$ . A conceptual design is presented and explained with the aid of a detailed block diagram and some of the implementation and system aspects of this hybrid candidate system are discussed.

### 7.1 SELECTION OF A CANDIDATE SYSTEM FOR DETAILED CONCEPTUAL DESIGN

The criteria for selecting one of the candidate systems presented in paragraph 6.0 for detailed conceptual design include

- cost/implementation complexity
- flexibility
- processing capability
  - MTF correction or image enhancement
  - BWC
  - Decimation/Interpolation
  - REARCS
  - throughput rate
- reliability
- performance

From the discussion and tradeoff matrix of paragraph 5.0 it was concluded that both the hybrid DCT/DPCM

technique and the two-dimensional DCT technique give superior BWC performance but at the expense of relatively high complexity when conventional single-point scanning is used. It was demonstrated in paragraph 6.0 that hybrid DCT/DPCM and two-dimensional DCT implementations are considerably simplified when array scanning is used in conjunction with CCD's.

The two-dimensional FIR filtering computer simulation experiments reported in paragraph 4.0 showed that fairly large FIR filters, e.g., at least larger than  $15 \times 15$  and perhaps as large as  $31 \times 31$ , are required in order to obtain suitable decimation/interpolation filtering performance. While this requires very large digital memories and very fast digital arithmetic capabilities for single-point scanning systems, the use of array scanning allows both the storage and computation required for two-dimensional (or one-dimensional) filtering to be performed using CCD's operating upon analog sampled-data pixel values.

By comparing the complexity of the various ASITS implementations presented in paragraph 6.0 it is clear that using CCD technology is the key to implementing a high performance, low complexity, high speed, flexible image transmission system. While both the two-dimensional DCT and the hybrid DCT/DPCM candidate systems are attractive in terms of the selection criteria the superior subjective and objective image quality of the hybrid technique and its potential for higher image throughput rate make it the logical choice for the detailed conceptual design.

## 7.2

### THE HYBRID MIXED-DOMAIN APPROACH TO SPATIAL RESOLUTION CONTROL

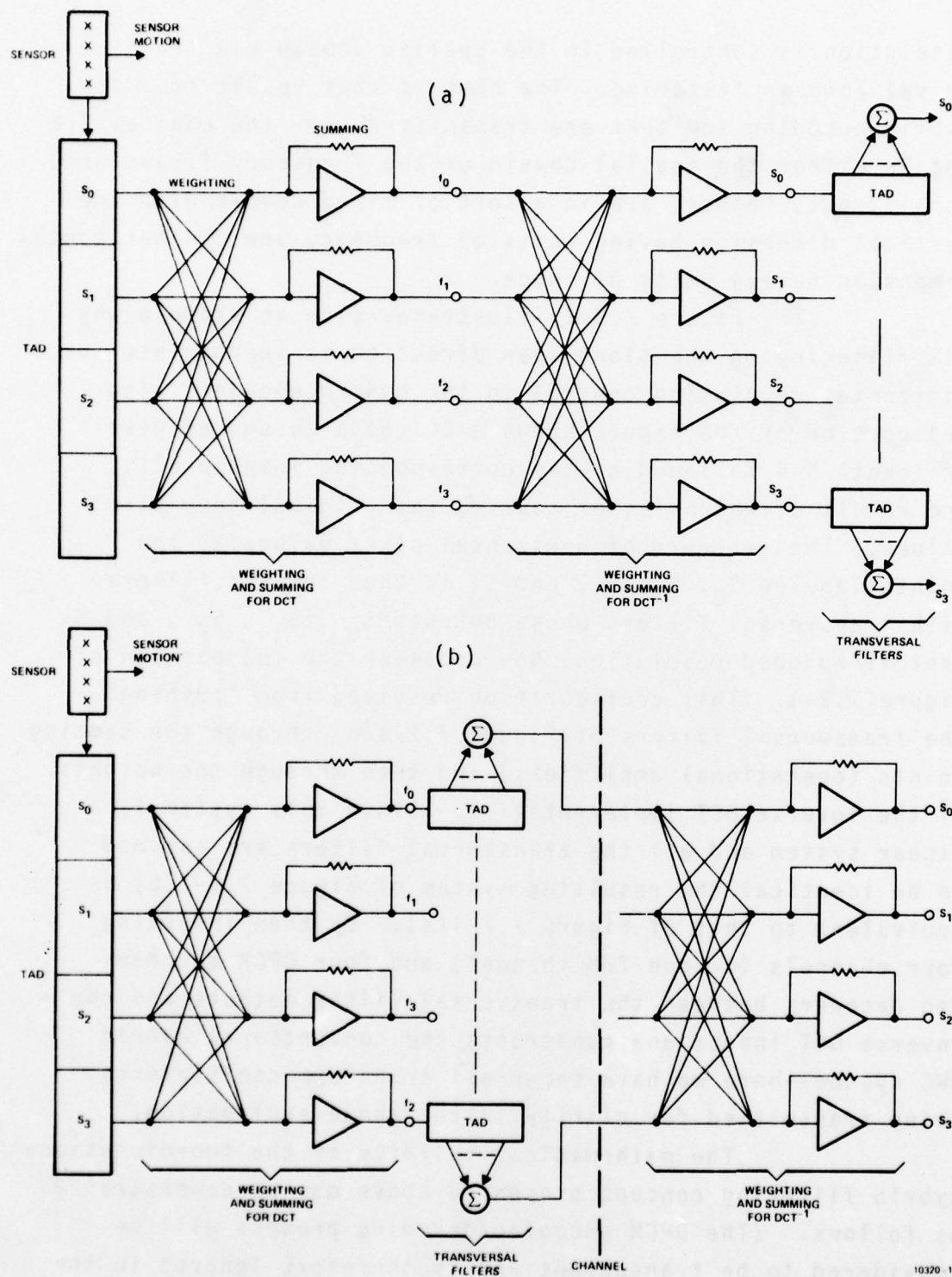
Recall from the discussion of Section 6.6 that for the hybrid BWC technique the vertical spatial resolution is controlled in the frequency domain by eliminating coefficients of the vertical DCT while the horizontal spatial

resolution is controlled in the spatial domain via transversal lowpass filtering. The numbers that result from the hybrid encoding and that are transmitted over the channel are not in either the spatial domain or the frequency (transform) domain, but, rather, are in a sort of mixed domain with the vertical dimension having units of frequency and the horizontal dimension having units of space.

Figure 7.2-1 illustrates with an example why FIR filtering in the along-scan direction is appropriate for horizontal resolution control in the hybrid encoder. The (a) portion of the figure shows a DCT of a column of pixels of length  $N=4$  followed by the corresponding inverse DCT, the result of which is, of course, the original four pixel values. The sequence of horizontal pixel values at the points labeled  $S_0$ ,  $S_1$ ,  $S_2$ , and  $S_3$  is then lowpass filtered with transversal filters whose outputs  $S'_0$ ,  $S'_1$ ,  $S'_2$ , and  $S'_3$  exhibit reduced resolution. Now consider the (b) portion of Figure 7.2-1. This configuration resulted from "pushing" the transversal filters of Figure 7.2-1(a) through the summing points (operational amplifiers) and then through the weights of the inverse DCT implementation. Since this system is a linear system and all the transversal filters are assumed to be identical the resulting system of Figure 7.2-1(b) is equivalent to that of Figure 7.2-1(a). By then inserting four channels (or one TDM channel) and four DPCM encoders and decoders between the transversal filter outputs and the inverse DCT inputs one constructs the conventional hybrid BWC system--here we have shown all transform coefficients being transmitted for clarity in the above explanation.

The mathematical validity of the two-dimensional hybrid filtering concept presented above may be demonstrated as follows. (The DPCM encoding/decoding process will be considered to be transparent and is therefore ignored in the following). The notation to be used in the derivation is





defined in Figure 7.2-2. This derivation is for the case in which the one-dimensional transformation is the Discrete Fourier Transform (DFT), although the conclusion is equally valid for the DCT. Using the DFT in this derivation facilitates the interpretation of the transform coefficients as being conventional spatial frequency components.

The sequence  $\{X(m,n)\}$  represents the image pixel values whose spatial location is indexed by row  $m$  and column  $n$ . Performing the DFT vertically (along columns) yields the transform coefficients

$$\bar{X}_k(n) = \sum_{m=0}^{N-1} X(m,n) \exp(j2\pi mk/N) \quad (7-1)$$

Filtering the resulting two-dimensional (mixed domain) sequence in the vertical direction is accomplished by merely scaling the vertical transform coefficients by the constants  $a_k$ , whereas filtering the sequence in the horizontal direction is accomplished by using a FIR transversal filter to perform the following convolution:

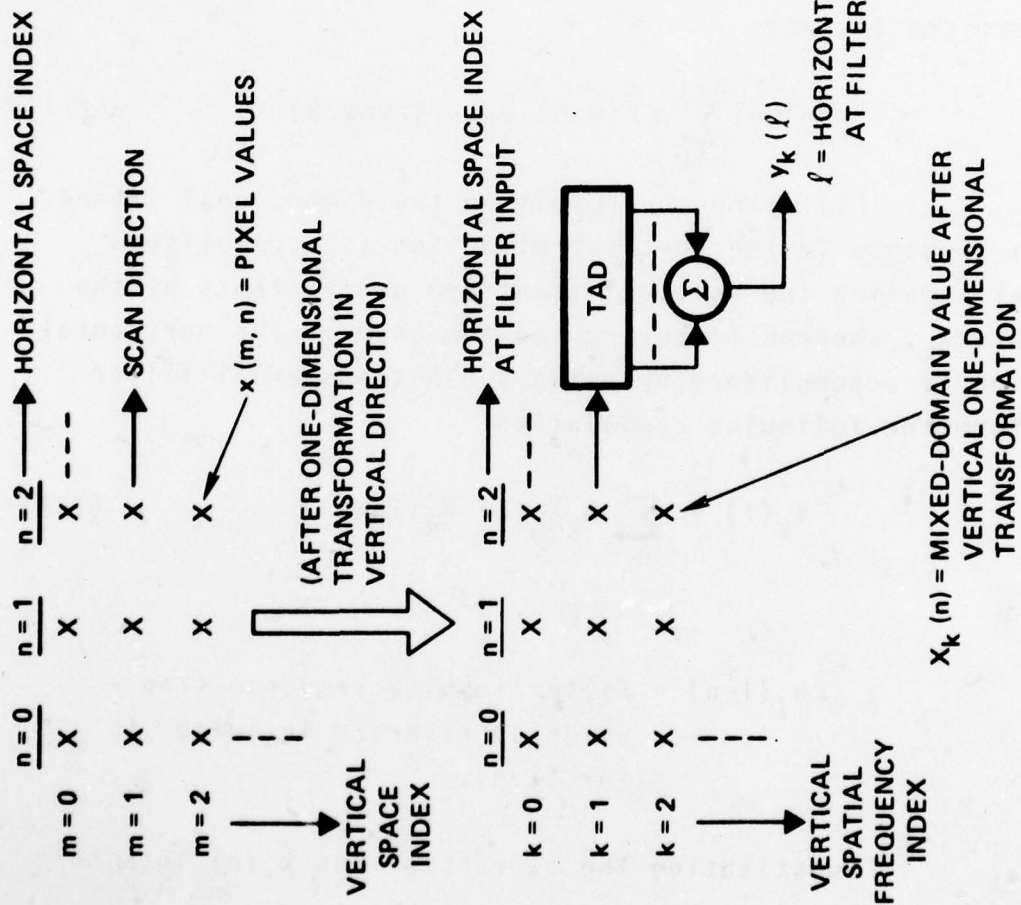
$$Y_k(l) = \sum_n a_k \bar{X}_k(n) h_1(l-n) \quad (7-2)$$

where

$h_1(l-n)$  = filter impulse response (tap weights) reversed in space (or time).

Substituting the expression for  $\bar{X}_k(n)$  into this expression one obtains:

$$Y_k(l) = a_k \sum_n h_1(l-n) \sum_{m=0}^{N-1} X(m,n) \exp(j2\pi mk/N) \quad (7-3)$$



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Figure 7.2-2  
Definition of Notation to be Used in Hybrid Filtering Derivation

We now take the inverse vertical DFT in order to return to the fully spatial domain from the mixed domain; this corresponds to the function performed by the receiver in the hybrid BWC technique. From the definition of the inverse DFT

$$Z_s(l) = \frac{1}{N} \sum_{k=0}^{N-1} Y_k(l) \exp(-j2\pi sk/N) \quad (7-4)$$

where  $l$  is the horizontal space index,  $s$  is the vertical space index at the receiver, and  $Z_s(l)$  is the two-dimensional sequence of the reconstructed image pixel values. Substituting for  $Y_k(l)$  in the above definition yields

$$Z_s(l) = \frac{1}{N} \sum_{k=0}^{N-1} \exp(-j2\pi sk/N) \sum_n A_k \bar{X}_k(n) h_1(l-n) \quad (7-5)$$

and substituting for  $\bar{X}_k(n)$  gives

$$Z_s(l) = \sum_{k=0}^{N-1} A_k \exp(-j2\pi sk/N) \sum_n h_1(l-n) \sum_{m=0}^{N-1} X(m,n) \cdot \exp(+j2\pi mk/N) \quad (7-6)$$

$$Z_s(l) = \sum_n h_1(l-n) \sum_{m=0}^{N-1} X(m,n) \left[ \frac{1}{N} \sum_{k=0}^{N-1} A_k \exp(j2\pi k(m-s)/N) \right] \quad (7-7)$$

Recognizing that the bracketed term is a function of  $(s-m)$ , which function we will designate as  $h_2(s-m)$ , the reconstructed two-dimensional pixel sequence may be written as

$$Z_s(l) = \sum_n \sum_{m=0}^{N-1} X(m,n) h_2(s-m) h_1(l-n) \quad (7-8)$$



But this last expression is simply the convolution of the original two-dimensional image pixel sequence with a separable FIR filter impulse response

$$h_{TOT}(s,l) = h_1(l)h_2(s) \quad (7-9)$$

This result demonstrates that the reconstructed image at the output of the hybrid decoder has been two-dimensionally filtered by a separable frequency transfer function.

Recall from paragraph 4.0 that using separable two-dimensional FIR filters to effect MTF correction (or image enhancement) suffers from the disadvantage that the two-dimensional MTF correction filter frequency response exhibits cusps or peaks at the corners of the transfer function. This is a result of the fact that for separable filters the two-dimensional frequency transfer function is the product of a one-dimensional vertical filter and a one-dimensional horizontal filter. Thus, for example, if a one-dimensional MTF correction filter with 10 dB gain at  $f_c$  is used for both the vertical and horizontal filtering, then the resulting separable two-dimensional MTF correction filter has 20 dB gain at  $(f_h = f_c, f_v = f_c)$  and only 10 dB gain at  $(f_h = 0, f_v = f_c)$  and  $(f_h = f_c, f_v = 0)$ . This is an undesirable result since some frequencies are emphasized more than others. Since the proposed hybrid technique implements separable rather than nonseparable FIR filters, the ASITS MTF correction will not be uniform within the passband of the system two-dimensional transfer function. The separable MTF correction computer simulation experiments reported in paragraph 4.0 indicate that this is an acceptable price to pay for the much lower implementation complexity of separable filters compared to nonseparable filters.

DETAILED BLOCK DIAGRAM FOR CONCEPTUAL DESIGN

Figure 7.3-1 shows a detailed block diagram presenting the hybrid technique proposed for the ASITS. The ASITS pictured in Figure 7.3-1 is capable of

- MTF correction/image enhancement
- REARCS operation
- BWCR at 4:1, 2:1, or 1:1 using the hybrid technique
- Resolution reduction of 4:1, 2:1, or 1:1
- Decimation and interpolation by factors of 2 and 4, in conjunction with two-dimensional separable lowpass filtering

These ASITS capabilities are discussed below in the explanation of the block diagram of Figure 7.3-1.

The hybrid ASITS block diagram of Figure 7.3-1 shows a conceptual design that makes liberal use of Reticon R5602 CCD's (designated by dashed line enclosures in the figure). These 64-tap mask programmable CCD's are used to perform three distinct functions in this implementation: DCT basis function correlation, separable horizontal lowpass filtering at intermediate and low resolution with integrated MTF correction or image enhancement, and image enhancement or MTF correction at maximum resolution where no lowpass filtering is required.

The linear array sensing technique selected for this conceptual design is the AOB, dithered raster, galvanometer-based approach discussed in detail in Paragraph 2.0. This approach is based upon well-developed technologies and is regarded as low-risk. Extensive trade-off analyses

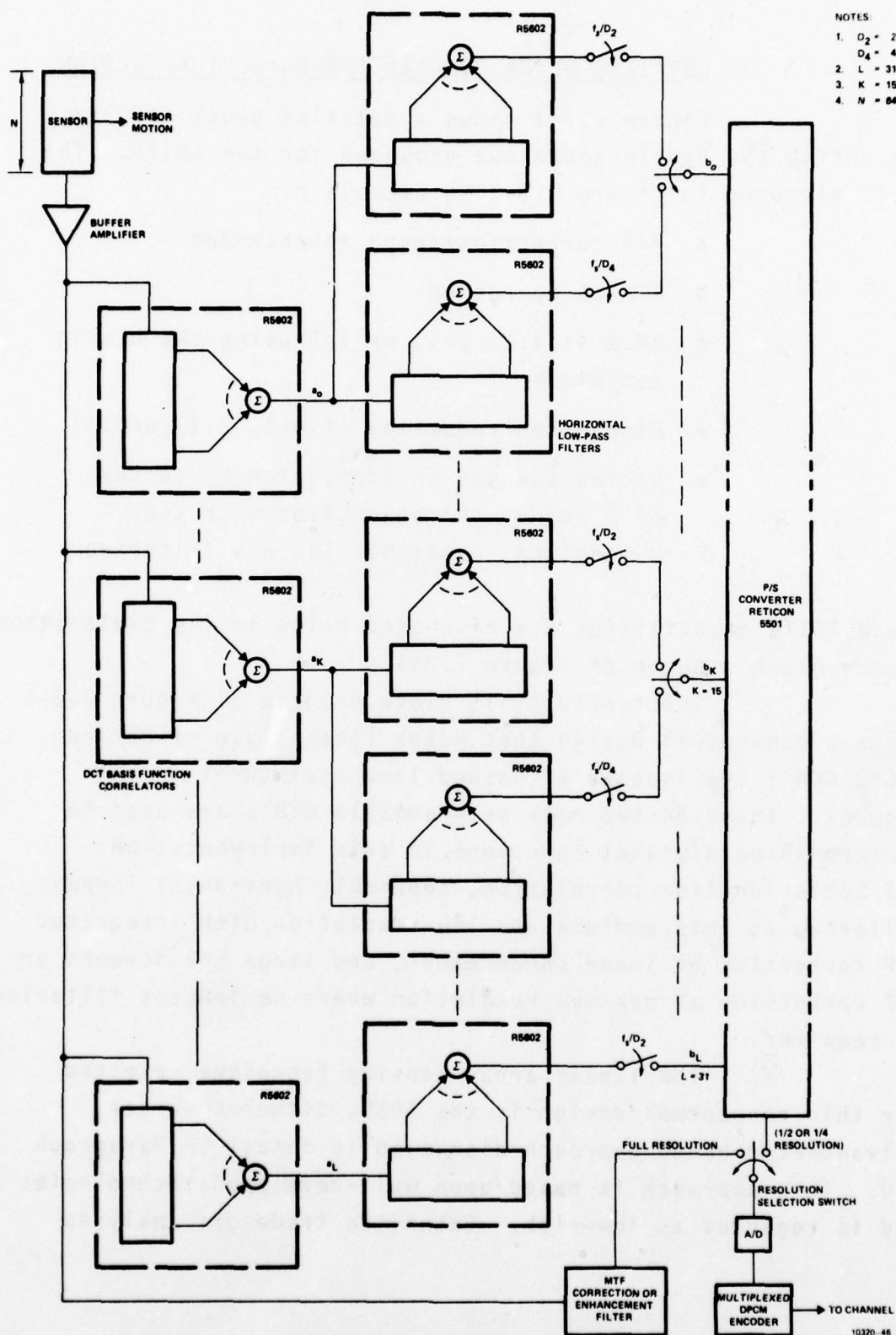


Figure 7.3-1. ASITS Hybrid Transformer Block Diagram

for this approach were presented in paragraph 2.0. A 64 element sensor is required in order to fully utilize the processing capability of the Reticon R5602 CCD's used to perform the hybrid transform.

The ASITS operation is as follows. The sensor array serially outputs 64 pixel values corresponding to a vertical column within the image being scanned; the vertical spacing between pixels is  $1/384 \text{ ppi} = 0.002604 \text{ inches}$ . These 64 pixel values are buffered and shifted simultaneously into 32 ( $L=31$ ) R5602 CCD's. Each of these R5602 CCD's is mask programmed with one of the first 32 DCT basis functions of the 64-point DCT. These basis functions are correlated with the 64 pixel values that are stored in the R5602 CCD's so that the CCD outputs are the DCT coefficients  $A_0$  through  $A_{31}$ . These thirty-two DCT coefficients are required in order to obtain one-half the full vertical resolution, since the DCT is of length 64. For the one-quarter resolution case, only 16 of the 64 DCT coefficients need be computed ( $K=15$  in Figure 7.3-1). A new set of DCT coefficients  $A_0$  through  $A_{31}$  is computed corresponding to each new set of 64 pixel values output by the sensor array; the sensor array outputs these 64 pixel values by horizontally advancing a distance equal to the pixel separation,  $1/384 \text{ ppi} = .002604 \text{ inches}$ . Each of the coefficient sequences is convolved with a sampled filter impulse response in order to provide horizontal lowpass filtering and MTF correction prior to along-scan decimation. Since both intermediate (192 ppi) and low resolution (96 ppi) modes must be provided, two sets of horizontal lowpass filters are required for the first 16 DCT coefficients with cutoff frequencies of  $f_s/4$  and  $f_s/8$ , respectively. The appropriate filter outputs are selected for parallel-to-serial conversion, depending upon whether 192 ppi or 96 ppi resolution is desired.



The parallel-to-serial converter shown in Figure 7.3-1 is a 32-stage Reticon R5501, which is a recently introduced Reticon product. If full resolution (384 lpi) is required then both the vertical DCT and the horizontal lowpass filtering operations are bypassed and the resolution selection switch of Figure 7.3-1 is switched to the full resolution position. In this situation the sensor data passes through the MTF correction or image enhancement filters indicated in Figure 7.3-1. The implementation of these filters is shown in Figure 7.3-2. The 7x7 nonseparable MTF correction filter utilizes a R5602 CCD with  $(64-49)=15$  tap weights programmed to zero. Recall that nonseparable MTF correction filters are more desirable (i.e., more flexible) than separable filters. Since at full resolution no stopband is required, low order filters should be adequate for MTF correction. Note that with 64 pixel values per column and a 7x7 FIR filter implemented in one linear CCD shift register, a clock enable (CE) signal is required so that only 7 of the pixel values comprising each column are actually shifted into the CCD. Although any 7 of the 64 pixel values could be selected, we will choose to read the 7 "topmost" pixel values of the 64 pixel column into the filter CCD's. This requires that the CE signal be inhibited during the time that the first 57 pixel values of each column are being read out of the sensor. Similarly, the CE signal must be inhibited during the first 61 sample times if image enhancement rather than MTF correction is desired. Note from Figure 7.3-2 that a switch controls the selection of image enhancement or MTF correction. Recall from Section 4.2 that any desired degree of image enhancement can be obtained using a 3x3 FIR filter with appropriate tap weights. The 3x3 enhancement filter can utilize either a R5602 or a TAD-32A; the latter choice allows

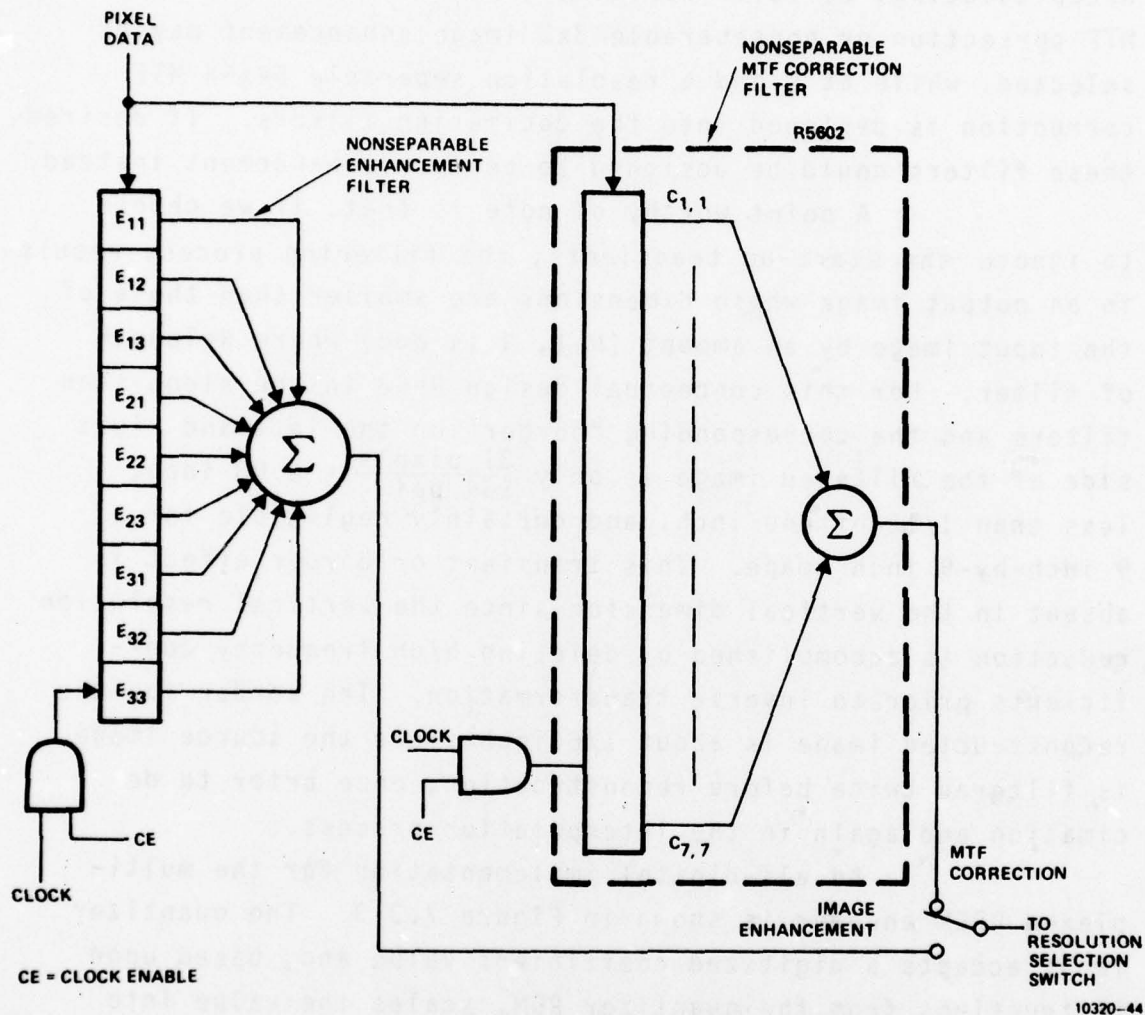


Figure 7.3-2  
Selectable Nonseparable MTF Correction  
and Image Enhancement Filters

the degree of image enhancement to be changed by simply adjusting 9 resistor values which set the FIR tap weights. Recapitulating, at full resolution, either nonseparable 7x7 MTF correction or nonseparable 3x3 image enhancement may be selected, while at  $\frac{1}{2}$  and  $\frac{1}{4}$  resolution separable 64x64 MTF correction is designed into the decimating filters. If desired, these filters could be designed to perform enhancement instead.

A point worthy of note is that, if we choose to ignore the start-up transients, the filtering process results in an output image whose dimensions are smaller than those of the input image by an amount  $(N-1, N \text{ is odd})$  where  $N$ =length of filter. For this conceptual design  $N=63$  in the along-scan filters and the corresponding "border" on the left and right side of the filtered image is only  $\frac{31 \text{ pixels}}{384 \text{ ppi}} = 0.08 \text{ inch}$ , less than  $1/12$  of one inch, and certainly negligible for a 9 inch-by-9 inch image. This transient or border effect is absent in the vertical dimension since the vertical resolution reduction is accomplished by deleting high frequency coefficients prior to inverse transformation. The border for the reconstructed image is about  $1/6$  inch since the source image is filtered twice before reconstruction, once prior to decimation and again in the interpolation process.

An all-digital implementation for the multiplexed DPCM encoder is shown in Figure 7.3-3. The quantizer input accepts a digitized coefficient value and, based upon instructions from the quantizer ROM, scales the value into the proper quantizer range via the appropriate number of shifts and sets the correct number of LSB's to zero; the forced-zero LSB's out of the quantizer are necessary for the predictor arithmetic, but are not transmitted over the channel. The proper scaling factor and quantizer accuracy are accessed in the quantizer ROM by forming an address using the BWCR CMD

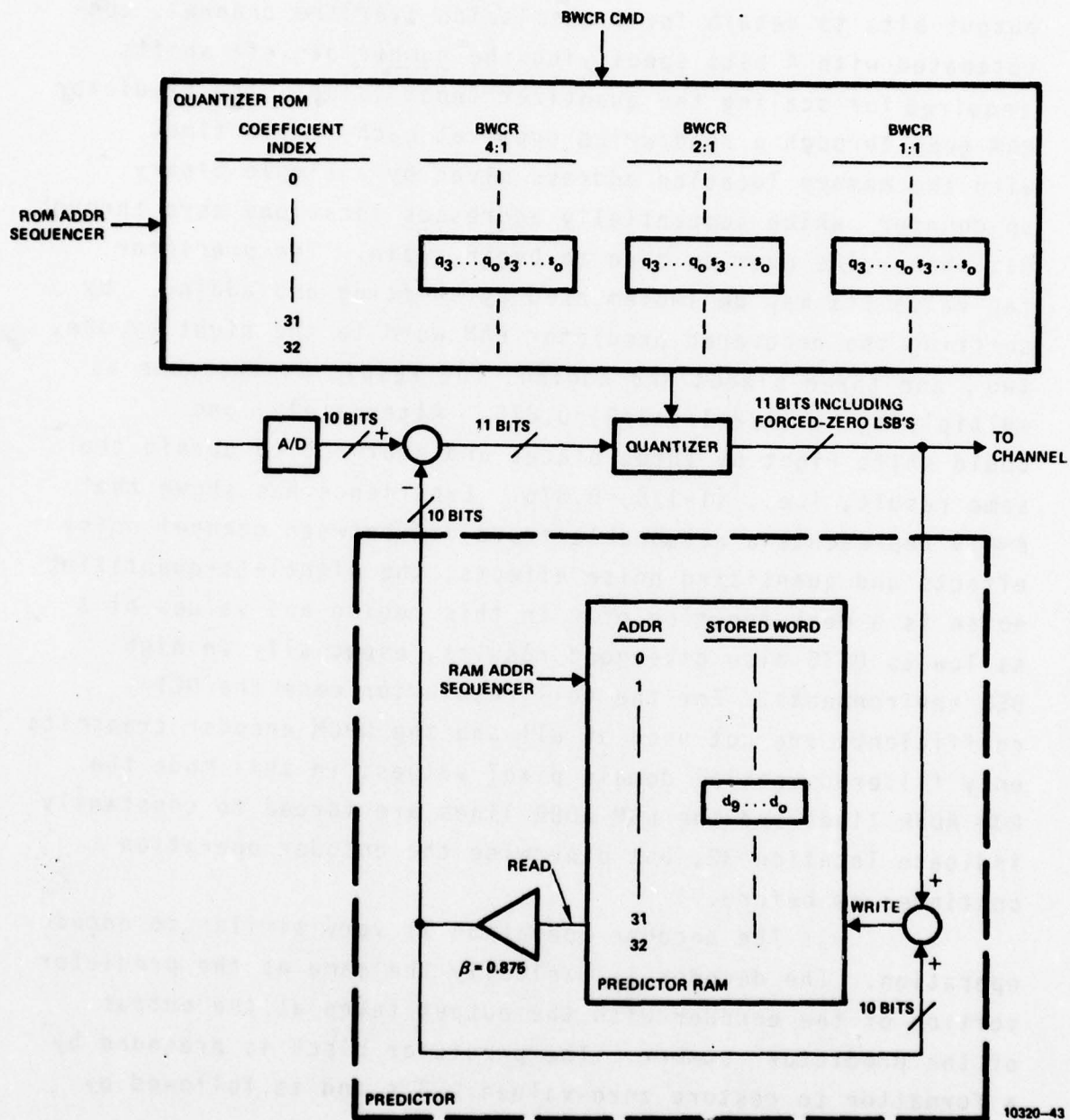


Figure 7.3-3  
All-Digital Implementation of Multiplexed DPCM Encoder



and the coefficient index (zero through 31). The ROM quantizer word consists of 4 bits designating the number of quantizer output bits to retain for transmission over the channel, concatenated with 4 bits specifying the number of left shifts required for scaling the quantizer input value. The predictor RAM goes through a read/write cycle at each sample time, with the memory location address given by a simple binary up-counter which sequentially addresses locations zero through 31, then rolls over to zero to begin again. The predictor tap value  $\beta < 1$  may be implemented by shifting and adding. By shifting the recovered predictor RAM word to the right by one, two, and three places and adding, the result is the same as multiplying by  $(1/2 + 1/4 + 1/8) = 0.875$ . Alternately, one could shift right by three places and subtract to obtain the same result, i.e.,  $(1 - 1/8) = 0.875$ . Experience has shown that  $\beta \approx 0.9$  represents a reasonable trade-off between channel noise effects and quantizing noise effects; the signal-to-quantizing noise is a weak function of  $\beta$  in this region and values of  $\beta$  as low as 0.75 also give good results, especially in high BER environments. For the full resolution case the DCT coefficients are not used at all and the DPCM encoder transmits only filtered spatial domain pixel values; in this mode the ROM ADDR lines and the RAM ADDR lines are forced to constantly indicate location 32, but otherwise the encoder operation continues as before.

The decoder operation is very similar to encoder operation. The decoder is basically the same as the predictor portion of the encoder with the output taken at the output of the predictor summer. The predictor block is preceded by a formatter to restore zero-valued LSB's and is followed by an inverse scaler.

The hybrid ASITS receiver is shown in Figure 7.3-4. The sequence of operations is the reverse of those at the hybrid transmitter. If full resolution mode is not selected the DCT coefficients from the output of the multiplexed DPCM decoder are inverse transformed to obtain 64 vertical pixel values. The horizontal sequence of pixel values that results from successive inverse transformations is then interpolated by a factor of two or four. The interpolated pixel values are stored in two Reticon 5501 CCD's P/S converters, whose serial outputs are alternately connected to the linear array recorder. This arrangement is necessary in order to synthesize a 64-stage P/S converter from two 32-stage P/S converters. If the full resolution mode is selected, then the buffered DPCM decoder output is read directly into the linear array recorder.

As in the hybrid transmitter implementation, liberal use is made of the mask-programmable 64-stage Reticon 5602 CCD. As mentioned earlier, the multiplexed DPCM decoder implementation is similar to that of the encoder shown in Figure 7.3-3 and is not discussed further. The hybrid ASITS receiver is of higher complexity than the hybrid ASITS transmitter in that twice as many R5602 CCD's are required for the inverse DCT. This is because only 32 of the 64 DCT coefficients need be computed at the transmitter (corresponding to one-half resolution) whereas all 64 spatial domain pixels must be computed via the inverse DCT at the hybrid receiver. Each of the 64-stage  $(DCT)^{-1}$  basis function correlators shown in Figure 7.3-4 is loaded with 32 DCT coefficients and 32 zeros for the one-half resolution case (or 16 coefficients and 48 zeros for the one-quarter resolution case) in order to perform the inverse DCT. The receiver complexity is further increased by virtue of the fact that two interpolating filters are

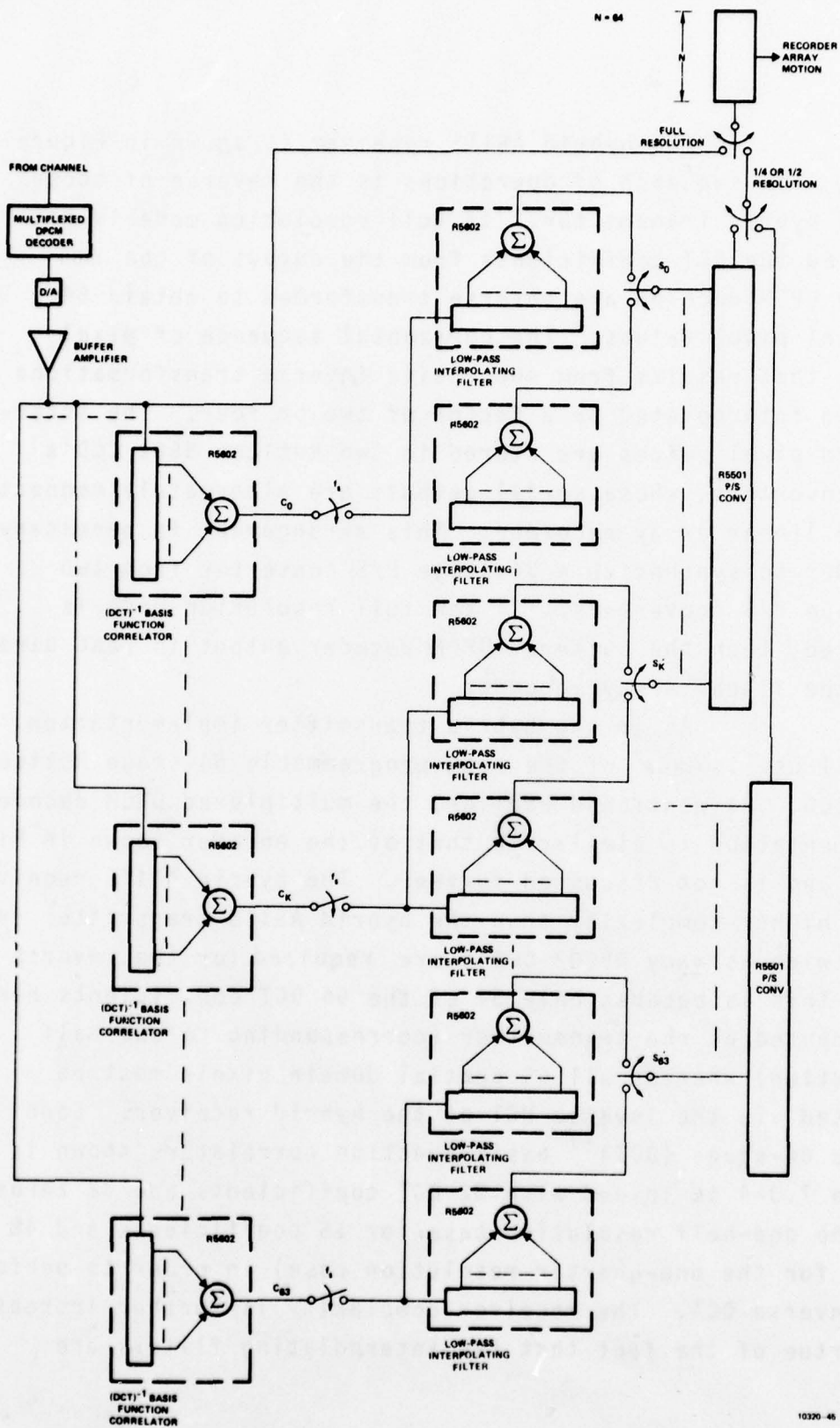


Figure 7.3-4. ASITS Hybrid Receiver Block Diagram

required for each of the 64 inverse DCT basis function correlators. Since the DCT is its own inverse, the first 32  $(\text{DCT})^{-1}$  basis function correlators are identical to the 32 DCT basis function correlators used in the hybrid transmitter. In addition, all the quarter-bandwidth decimating and interpolating filters in the system are identical, as are all the half-bandwidth decimating and interpolating filters.

The linear array recording technique selected for this conceptual design is the AOBD scanned recording approach discussed in some detail in paragraph 3.0. Of the several array recording approaches evaluated in paragraph 3.0, the AOBD approach is the lowest in cost, and its  $N \leq 1500$  element, 75 M pixels/sec capability insures that it can easily meet the ASITS requirements we have specified. However, it was pointed out in paragraph 3.0 that it is difficult to change scanning speeds with this technique; this is a factor of importance discussed in Section 7.4, System Considerations and Trade-offs, which follows.

#### 7.4

#### SYSTEM CONSIDERATIONS AND TRADE-OFFS

Briefly stated, the ASITS system parameters are determined by the requirement to transmit a 9 inch-by-9 inch image quantized to four bits per pixel within ten minutes over a 4.8 Kbps channel. For the purpose of defining maximum pixel rates, it will be assumed that this image will have 12 square inches transmitted at a resolution of 384 ppi with the remaining 69 square inches transmitted at a resolution of 96 ppi. Simple calculations show that a bandwidth compression ratio of four-to-one throughout the image is adequate to meet the above requirement.



It is also desirable to be able to select a resolution of 192 ppi and a BWCR of 2:1 or 1:1 wherever desired within the image; of course, the 10 minute transmission time requirement will not be met under these circumstances. It should be pointed out that specifying a 4:1 BWCR for both the target region and the background region would probably not be typical of ASITS operation since 4:1 BWCR in the target region is equivalent to 1-tap delta modulation and would not yield the image quality desired for the target region; the specification of 4:1 BWCR in both regions in the example above simply serves to determine the pixel rate required.

Before computing the required pixel rates at the sensor, recorder, and CCD processors, we present several possible bit assignments for the hybrid technique. In discussing the bit assignments and their relationships to the BWCR we will treat resolution reduction and BWCR as separate processes in keeping with the spirit of the REARCS definitions of these terms. Since we assume a 64-point DCT, this means that 32 coefficients will be retained for half resolution and 16 coefficients will be retained for quarter-resolution. The horizontal resolution is controlled by decimation filters in each case. The bit assignment over the retained coefficients then determines the BWCR as we have defined it in this report. For example, if we retain only 16 of the 64 coefficients, we call this the quarter resolution case. Furthermore, if we then choose a bit assignment for these 16 retained coefficients that has a total of, say, 32 bits, then this corresponds to a BWCR of  $32 \text{ bits} / 16 \text{ coefficients} = 2 \text{ bits/pixel}$  at one quarter resolution, where these pixels are low resolution pixels prior to interpolation. This definition of BWCR is in contrast to the one commonly found in the literature which includes the unretained coefficients in the BWCR computation; however,

our definition is consistent with the REARCS definitions of BWC and image resolution. The BWCR is defined in terms of the pixels that remain after resolution reduction has taken place. In Table 7.4-1 we present for illustration purposes some bit assignments associated with BWCR's of 4:1, 2:1, and 1:1 for both the half-resolution and quarter resolution cases. These bit assignments are estimates based upon some prior experience with the hybrid technique and have not been optimized for this application. However, they do serve to illustrate the general nature of hybrid transform bit assignments. Note that for the 1 bit/pixel case some of the higher coefficients are set to zero, which further decreases the real available resolution below one-quarter or one half; this is unavoidable since assigning one bit to each coefficients is clearly sub-optimum. Assigning more bits to the lower coefficients results in better image quality, and for the 1 bit/pixel case this necessitates assigning zero bits to some of the higher "retained" coefficients.

We now examine the required swath scanning rate (inches/second) and the associated pixel scanning rate (pixels/second) at the sensor output. This will allow us to determine whether the pixel rate capabilities of the sensor, recorder, and CCD processors are exceeded when the system throughput requirement stated earlier is imposed. Figure 7.4-1 depicts the scanning action at either the sensor or the recorder and shows that each sweep of the 64-element sensor (or recorder) across the image covers an area  $1/6$  inch by 9 inches, hereafter called a swath. In the background (low resolution) region of the image the sensor advances 64 lines at a time in non-overlapping swaths. However, in the target (full-resolution) region the sensor advances only one line at a time in order to accomplish the two-dimensional MTF correction filtering--the DCT is not performed. In those portions of the image

TABLE 7.4-1  
HYBRID TRANSFORM BIT ASSIGNMENT VERSUS  
BWCR AND RESOLUTION SELECTION

<u>Quarter Resolution Case</u>				<u>Half Resolution Case</u>		
	BWCR			BWCR		
<u>Coefficient</u>	<u>4:1</u>	<u>2:1</u>	<u>1:1</u>	<u>4:1</u>	<u>2:1</u>	<u>1:1</u>
C <sub>0</sub>	4	7	9	3	8	10
C <sub>1</sub>	2	5	8	2	6	9
C <sub>2</sub>	1	4	7	2	5	9
C <sub>3</sub>	↓	3	7	1	4	9
C <sub>4</sub>		2	7	↓	3	8
C <sub>5</sub>		1	6		3	8
C <sub>6</sub>		↓	5		3	7
C <sub>7</sub>			3		2	6
C <sub>8</sub>			2		2	6
C <sub>9</sub>			2		2	5
C <sub>10</sub>	↓		2		2	5
C <sub>11</sub>	1		2		2	4
C <sub>12</sub>	0		1		2	4
C <sub>13</sub>	↓		1		2	4
C <sub>14</sub>		↓	1		1	4
C <sub>15</sub>		1	1		↓	3
C <sub>16</sub>		0	0			3
C <sub>17</sub>		↓				3
C <sub>18</sub>						3
C <sub>19</sub>						2
C <sub>20</sub>						2
C <sub>21</sub>						2
C <sub>22</sub>				↓	↓	2
C <sub>23</sub>				1	1	2
C <sub>24</sub>				0		1
C <sub>31</sub>				↓		1
C <sub>32</sub>						0
C <sub>63</sub>						↓

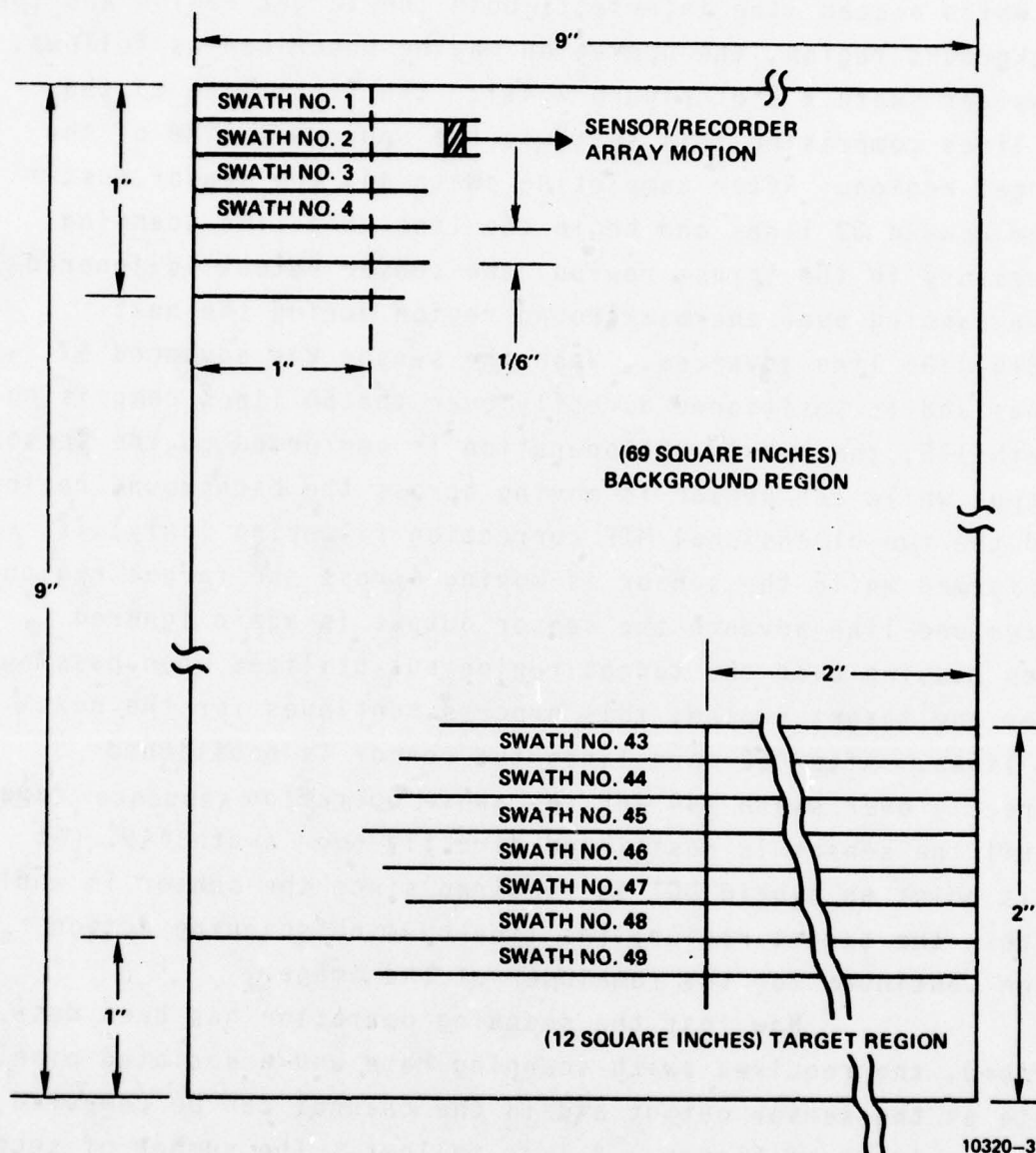


Figure 7.4-1  
The Nature of the  
Linear Array Scanning Action



in which a scan line intersects both the target region and the background region, the operation may be described as follows. Consider swath #43 of Figure 7.4-1: the first line of the 64 lines comprising this swath is the uppermost line of the target region. After completing swath #43 the sensor must move upward 32 lines and begin the line-at-a-time scanning necessary in the target region; the sensor output is ignored when passing over the background region during the next  $(32+64)=96$  line advances. When the sensor has advanced 97 lines and is positioned directly over the 64 lines comprising swath #45, the hybrid DCT operation is performed on the sensor output while the sensor is moving across the background region and the two-dimensional MTF correction filtering (only) is performed while the sensor is moving across the target region. After one line advance the sensor output is again ignored when passing over the target region but utilized when passing over the target region; this process continues for the next 62 lines. After 62 more lines the sensor is positioned directly over swath #44 and the above operation sequence repeats until the sensor is positioned directly over swath #49. At this point no hybrid DCT is required since the sensor is wholly within the target region; the line-by-line scanning action then continues for the remainder of the image.

Now that the scanning operation has been described, the required swath scanning rate and associated pixel rate at the sensor output and in the channel can be computed for the image of Figure 7.4-1 as follows. The number of seconds  $T_B$  required to scan the full width of 42 swaths at a constant swath scanning rate SR is

$$T_B = (9(\text{in/swath})/\text{SR}(\frac{\text{in}}{\text{sec}})) \times 42 \text{ swaths} = \frac{378}{\text{SR}} \text{ sec} \quad (7-10)$$

In the target region the sensor array advances only one scan line at a time rather than 64 scan lines at a time. The number of seconds  $T_T$  required to scan the 2 in x 384 lpi = 768 lines passing through the target region is:

$$T_T = 9(\text{in/line})/\text{SR}(\text{in/sec}) = 768 \text{ lines} = \frac{6912}{\text{SR}} \text{ sec.} \quad (7-11)$$

Note that the 32 lines of swath #42 that must be rescanned exactly compensate the last 32 lines of the bottom swath that are not scanned due to border effects. Now since  $T_T + T_B = 10$  min (600 sec) the required swath scanning rate may be computed as follows:

$$T_B + T_T = 600 \text{ sec} = (378/\text{SR} + 6912/\text{SR}) \text{ sec.} \quad (7-12)$$

Then  $\text{SR} = 12.15$  (inches/sec) and the time required to scan one swath is  $9 \text{ inches} / 12.15 \text{ inches/sec} = 0.74074 \text{ sec}$ . The corresponding pixel rate  $\text{PR}_S$  at the sensor output is given by

$$\begin{aligned} \text{PR}_S &= (64 \text{ pixels/column}) \times (384 \text{ columns/inch}) \times (\text{SR} \frac{\text{inches}}{\text{sec}}) \\ &= 18,186 \text{ pixels/sec} \end{aligned} \quad (7-13)$$

This sensor pixel rate is well within the capability of the sensor array, recording array, and CCD image processing devices proposed for this conceptual design implementation.

There is a disadvantage to using a constant swath scanning rate, as was assumed for the above calculations: the pixel rate (and therefore the bit rate) in the channel is four times faster when scanning the background region than when scanning the target region, which implies the need for a

channel buffer if a constant channel data rate is desired or required. This may be demonstrated as follows. Recall that 4:1 BWCR implies 1 bit/pixel in both target and background regions. In the target region the channel pixel (bit) rate may be computed as

$$\begin{aligned} PR_T &= (384 \text{ pixels/in})(9 \text{ in})/(0.74074 \text{ sec}) \\ &= 4,666 \text{ pixels/sec} = 4,666 \text{ bits/sec} \end{aligned} \quad (7-14)$$

In the background the along-scan pixels are decimated by four, but 16 of the 64 vertical DCT coefficients are retained and transmitted at an average rate of one bit/pixel for 4:1 BWCR; the channel bit rate in the background region is thus given by

$$\begin{aligned} PR_B &= \frac{(9 \frac{\text{in}}{\text{scan}})(16 \frac{\text{coefficients}}{\text{pixel column}})(384 \frac{\text{pixels}}{\text{in}} + 4 \text{ decimation})}{0.74074 \text{ sec/scan}} \\ &= 4 \times 4.666 = 18,662 \text{ bits/sec} \end{aligned} \quad (7-15)$$

If the swath scan rate SR is to be held constant the size of the required channel buffer is then bounded by  $B = (\text{differential bit rate}) \times (\text{total scan time for 42 contiguous background region swaths}) = (18,662 - 4800) \text{ bits/sec} \times (378/12.15) \text{ sec}$   
 $B = 575,796 \text{ bits.}$

The requirement for channel buffers could be eliminated by decreasing the swath scanning rate by a factor of four in the background regions so as to obtain a constant channel data rate. However, if the swath scanning rate is not speeded up (from 4.666 Kbps to 4.8 Kbps) in order to compensate, the 10 minute transmission time requirement will be exceeded.

Accounting for the fact that one swath may intersect both target and background regions, and allowing SR to change within a swath after crossing a region boundary we may compute the two new swath scanning rates required for both kinds of regions in order to keep the link full as follows.

Using

$$T_B = \frac{(9 \text{ in/swath}) \times (42 + (7/9)6) \text{ swaths}}{SR_B} \quad (7-16)$$

and

$$T_T = \frac{(9 \text{ in/line}) \times (384 + 2/9(384)) \text{ lines}}{SR_T} \quad (7-17)$$

and

$$600 \text{ sec} = T_B + T_T = \left( \frac{382.7}{SR_B} + \frac{4224}{SR_T} \right) \text{ sec} \quad (7-18)$$

we obtain one equation in two unknowns. Recall that with nonconstant SR (and thus constant channel data rate),

$$SR_B = 1/4 SR_T. \quad (7-19)$$

We may then write  $600 \text{ sec} = \frac{382.7}{SR_B} + \frac{4227}{4(SR_B)}$  and solve for

$$SR_B = 2.397 \text{ inches/sec}$$

$$SR_T = 4SR_B = 9.588 \text{ inches/sec}$$



The corresponding sensor pixel rates are then:

$$\begin{aligned}\text{background region: } PR &= (64 \text{ pixels/column})(384 \text{ columns/inch}) \times \\ &\quad (2.397 \text{ inches/sec}) \\ &= 58,909 \text{ pixels/sec}\end{aligned}$$

$$\begin{aligned}\text{target region: } PR &= (64 \text{ pixels/column})(384 \text{ columns/inch}) \times \\ &\quad (9.588 \text{ inches/sec}) \\ &= 235,634 \text{ pixels/sec}\end{aligned}$$

All of the computations thus far have assumed a 4.8 kbps link. Since sensor pixel rates are linearly related to SR we can easily tabulate the corresponding results for other data rates of interest. Table 7.4-2 shows these results for the first case we considered, in which the swath rate was held fixed throughout the image and channel buffers were required to accommodate the resulting nonconstant channel data rate. Note that required buffer size is independent of channel data rate. This is because the differential data rate  $\Delta d$  increases linearly with data rate  $d$  whereas total scan time decreases as  $1/d$ , and the required buffer size is the product of these two terms. Table 7.4-3 shows the corresponding results for the latter case we considered in which the SR is four times slower in the background region than in the target region in order to eliminate the channel buffer requirement.

TABLE 7.4-2  
SCANNER PIXEL RATES  
FOR THE CONSTANT SR SCANNER IMPLEMENTATION

<u>Available Channel Data Rate (kbps)</u>	<u>PR<sub>s</sub> Scanner Pixel Rate (kilopixels/sec)</u>
2.4	149.299
4.8	298.598
8.0	497.663
9.6	597.196
16	995.327
32	1,990.653

Note: required buffer size is 575,796 bits, independent of data rate

TABLE 7.4- 3  
SCANNER PIXEL RATES FOR THE  
NONCONSTANT SR SCANNER IMPLEMENTATION

<u>Available Channel Data Rate (kbps)</u>	<u>Scanner Pixel Rates (kilopixels/sec)</u>	
	<u>PR<sub>B</sub></u>	<u>PR<sub>T</sub></u>
2.4	29.455	117.817
4.8	58.909	235.634
8.0	98.182	392.723
9.6	117.818	471.268
16	196.363	785.447
32	392.727	1,570.893

In order to determine the capabilities of the selected array sensing technique when applied to this conceptual hybrid design for the ASITS, we must modify the results presented in Section 2.0, which were based on the assumption of a scan period  $T_{SS}=13.5$  milliseconds. We have shown that the scan periods required for the conceptual design presented here ( $N=64$ ) are as follows.

Constant SR approach:  $T_{SS}=0.74074$  seconds

Nonconstant SR approach:

background region:  $T_{SS}=9 \text{ inches}/2.397 \text{ inches/sec}=3.7547 \text{ sec}$

target region:  $T_{SS}=9 \text{ inches}/9.588 \text{ inches/sec}=0.9387 \text{ sec}$

Since  $N$  is directly proportional to  $T_{SS}$ , the scale for the vertical axis for  $N$  in the tradeoff summary curves of Figure 2.3.3.4 should be multiplied by the following scale factors in order to determine the maximum value of  $N$  applicable to the hybrid ASITS approach presented here:

Multiplicative Scale Factor

Constant SR approach:  $0.74074/.0135 = 54.9$

Nonconstant SR approach:

background:  $3.7547/.0135 = 278.1$

target:  $0.9387/.0135 = 69.5$

Choosing the worst case photodetector bandwidth of  $BW=1.0$  MHz for which  $N=10$  if  $T_{SS}=13.5$  msec implies that the maximum value of  $N$  for the 4.8 kbps channel data rate is

Constant SR approach:  $N=10 \times 54.9=549$

Nonconstant SR approach:

background:  $N=10 \times 278.1=2781$

target:  $N=10 \times 69.5=695$



Several notes are appropriate here about the feasibility of the very large array sizes just computed. The tradeoff being made is clearly a valid one, since it is simply a consideration of the transition between vertical-raster and horizontal-raster scanning; thus, ultimately, as many pixels as are scanned horizontally in a conventional system may be scanned vertically using a dithered raster technique, provided the relationship of the scan times can be adjusted accordingly. However, some caution must be used in applying these results to a specific scanner implementation. For the current case, with the system design as described in Paragraph 2.3.3, an additional constraint on array column height is posed by the detector size. In the absence of a more sophisticated system design, the long narrow detector can readily intercept light from only about 1/3 of an inch of document area. Thus the light budget (and hence SNR) tradeoffs would limit array size to about 128 pixels (at 384 pixels/inch). If we accept this limitation to a specific scanner design, then we may take the above SR calculations as evidence that there are no bandwidth problems anticipated for  $N \leq 128$ .

For the case at hand, i.e., for the  $N=64$  case corresponding to the proposed conceptual design, we are interested in determining whether the sensor pixel rates required at the various channel data rates can be supported by the selected scanning approach. Since the sensor pixel rate increases directly with  $N$  and inversely with swath scan period we may write

$$PR = (N/T_{SS})(9 \text{ in})(384 \text{ pixels/in}) \quad (7-20)$$

Referring to Figure 2.3.3.4 and choosing the (worstcase) photodetector bandwidth curve labeled  $BW=1.0 \text{ MHz}$  for which  $N = 10$  we may compute the maximum pixel rate

$$PR = \frac{10 \text{ pixels/column } 3456 \text{ columns/swath}}{13.5 \times 10^{-3} \text{ sec/swath}}$$

$$= 2.56 \times 10^6 \text{ pixels/sec}$$

The corresponding result for the BW = 2.6 MHz (best case) photodetector curve for which  $N=25$  is  $2\frac{1}{2}$  times larger or  $PR = 6.4 \times 10^6$  pixels/sec. Both of these sensor pixel rate capabilities exceed the sensor pixel rates required for the 32 Kbps link whether a constant SR approach or nonconstant SR is selected. Recall from Tables 7.4-2 and 7.4-3 that the 32 Kbps link requires maximum sensor pixel rates of  $1.990653 \times 10^6$  pixels/sec and  $1.570893 \times 10^6$  pixels/sec for the constant SR and nonconstant SR approaches, respectively. Thus, the sensor pixel rates dictated by the system parameters, viz.,  $N=64$  and 32 Kbps channel data rate, are well within the capability of the state of the art for the proposed array sensing technique. The array recording rate capability (75 Megapixels/sec) of the proposed array recording technique is more than an order of magnitude greater than the rate dictated by system requirements. In this section we have demonstrated that holding swath scanning rate SR constant implies a nonconstant channel data rate, and conversely, maintaining constant channel data rate implies a nonconstant swath scanning rate; the ratio of the two rates is 4:1 in either case. Conceptually at least, either of these approaches could be implemented without undue complexity. Since the source image target regions are designated and this information transmitted to the receiver prior to image transmission, the receiver knows when to expect data rate changes or when to change its swath scanning rate SR, depending upon whether channel data rate or swath scanning rate is held fixed.

An attractive alternative to the conceptual design just presented in this section is one in which the hybrid transform is performed for all three resolutions and BWCR's, not just for the quarter and half resolution cases. Such an approach would require that 64 DCT coefficients be computed at the transmitter rather than the 32 required for the approach presented in Figure 7.3-1. This alternative approach requires 32 additional Reticon R5602 CCD's for computing 32 additional DCT coefficients and one additional Reticon R5501 P/S converter and associated switch; however, the two Reticon R5602's associated with the image enhancement or MTF correction filters are no longer required. Thus, 30 additional R5602's and one R5501 are required for this alternative approach. This increased cost and implementation complexity must be evaluated in light of the higher image quality that results at high BWCR's when the hybrid BWC technique is used at full resolution. Since image quality is very subjective, this evaluation can only be accomplished by computer simulating this alternative system and comparing the reconstructed imagery to that of the system of Figure 7.3-1. Using the alternative approach also simplifies the scanning procedure in that no line ever need be scanned more than once, i.e., swaths do not overlap in the target region. The corresponding hybrid receiver for the alternative hybrid transmitter described above performs the inverse of the transmitter operations; its block diagram is not shown since it is a straightforward extension of the hybrid portion of the receiver of the conceptual design presented earlier.

Note that all of the tradeoff results have been based upon one particular geometry and ratio of image target area and background area; different geometries and area ratios will give different results. However, this choice is perhaps representative and does provide some insight into how the various system parameters are related.



It should be mentioned that, depending upon the power levels required, dangerous levels of coherent laser light may be present in the sensing or recording portions of the ASITS and due precautions should be taken when operating or servicing such equipment.

## 7.5 IMPLEMENTATION COST CONSIDERATIONS

The ASITS conceptual design presented in Section 7.4 makes liberal use of CCD technology at both the transmitter and the receiver to perform the two-dimensional filtering and BWC functions. The total number of Reticon R5602 64-stage transversal filters used at the transmitter is  $(3 \times 16 + 2 \times 16 + 2) = 82$ , while the number used at the receiver is  $3 \times 64 = 192$ , for a total of 274 per ASITS. Three Reticon R5501 P/S converters are required per ASITS, one at the transmitter and two at the receiver. Recall that the alternative ASITS that was briefly mentioned at the end of Section 7.4 required 30 additional R5602's and one additional R5501. The current prices for various quantities of these two Reticon CCD's are given in Table 7.5-1 below. The price structure of the Reticon TAD-32A is identical to that of the Reticon R5501.

The cost of the signal processing CCD's per complete ASITS may be easily computed and is given below in Table 7.5-2.

The support circuitry required to implement one R5602 filter is quite modest and consists of eight resistors, four  $10\mu\text{f}$  electrolytic capacitors, two  $0.1\mu\text{f}$  capacitors, 3 diodes, one zener diode, one 50K potentiometer, and one LF356 operational amplifier.

The Reticon R5602 is packaged in a standard 16-pin dual-inline-package (DIP) and requires only +15 volts for power and a single trigger input to control the internal



TABLE 7.5-1

## RETICON CCD PRICE STRUCTURE

RETICON DEVICE	<u>1-9</u>	<u>10-49</u>	<u>50-99</u>	<u>100-499</u>	<u>500-999</u>	<u>1000-2499</u>	<u>2500-4999</u>	<u>5000-9999</u>	<u>&gt;10,000</u>
R5602	\$50	\$40	\$30	\$25	\$17	\$14.50	\$12	\$10	\$8.5
R5501 or TAD-32A	\$48	\$35	\$25						

TABLE 7.5-2  
TOTAL COST OF R5602's AND R5501's  
PER ASITS VERSUS NUMBER OF ASITS FABRICATED

<u># of ASITS FABRICATED</u>	<u>1</u>	<u>5</u>	<u>10</u>	<u>20</u>	<u>50</u>
Conceptual Design	\$6994	\$4117	\$3393	\$2845	\$2404
Alternative Design	\$7792	\$4600	\$3788	\$3180	\$2724

sampling. Since the R5602 requires very little support circuitry, the required number of R5602's could be mounted on a relatively small number of printed circuit cards.

There are three basic approaches to implementing the hybrid ASITS processing functions:

- (1) utilize the architecture of Figure 7.3-1 and sampled-data analog hardware based upon CCD technology
- (2) utilize the architecture of Figure 7.3-1 but perform all of the storage and arithmetic functions using dedicated digital hardware
- (3) utilize a general purpose mini-computer and write the appropriate software to perform the hybrid algorithm

The capabilities of an ASITS implemented using either approach (1) or (2) are limited by the fixed architecture of the hardware, while the approach of (3) is completely general in that the software can be rewritten to perform any desired function. In general, system flexibility may be traded for system speed (image throughput rate). In comparing approaches (1) and (2), we find that the CCD implementation is conducive to relatively high image processing rates, whereas an all digital implementation requires more internal control and housekeeping functions, and is more constrained by inherent speed limitations.

For a relatively complex system such as the ASITS, the amount of software/firmware required for a microprocessor-based digital implementation can become an important contributor toward the overall system cost. For example, an all-digital, microprocessor-based, single-spot scanning hybrid

system requires several thousand lines of software of the industry-wide estimate for the cost of generating, debugging, and documenting software is on the order of \$25 per line.

In conclusion, it is clear that the CCD-oriented implementation of an ASITS is a cost effective low-risk approach that allows high image throughput rates while simultaneously performing sophisticated, computation-intensive image processing.



## 8.0 STUDY SUMMARY, CONCLUSIONS AND RECOMMENDATIONS

The conclusions of the various aspects of this study effort are documented throughout this final report. In this section we briefly review some of the major activities of the study and mention the corresponding conclusions. There is much useful data throughout the body of the report regarding pertinent tradeoffs, alternative architectures, and performance limits and degradations. Analysis, computer simulation, and some experimental work played major roles in the performance of this study. This section concludes with recommendations for further work that should be conducive to the ultimate realization of an ASITS making extensive use of CCD technology and novel applications of electro-optical technology.

### 8.1 STUDY SUMMARY AND CONCLUSIONS

This study has shown that an array scanning image transmission system is both feasible and advantageous when compared to conventional single-point scanned image transmission systems.

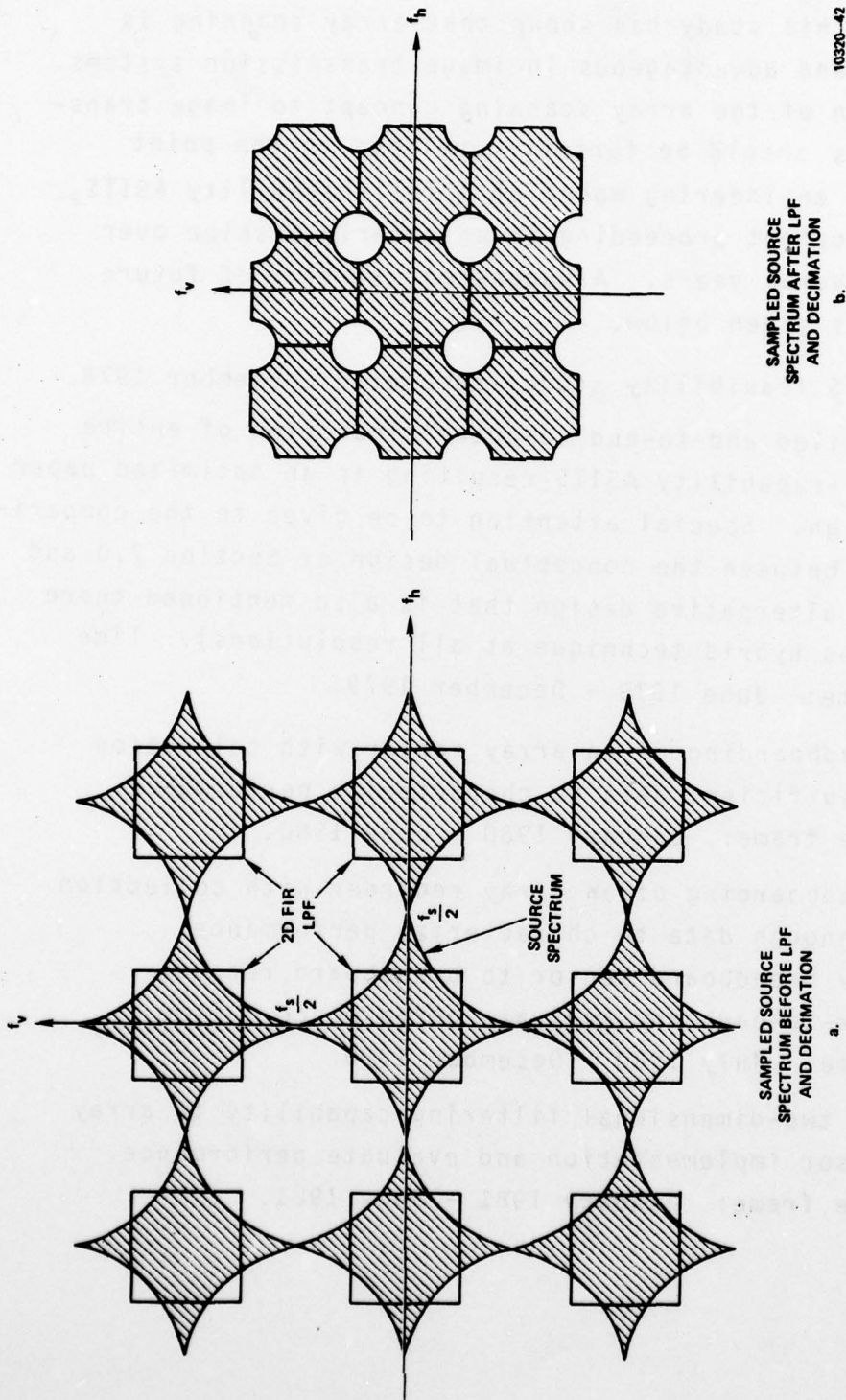
One major focus of the study was the evaluation of the risk and performance parameters of various array sensing and recording techniques. It was found that the current state of electro-optics technology does not support two-dimensional (matrix) array sensing or recording. However, the technology does support one-dimensional sensing and recording arrays capable of operating at useful pixel rates in image transmission systems. A one-dimensional linear sensing array  $N$  elements long can be used in conjunction with  $N^2$  memory locations to synthesize a two-dimensional sensing array. At the receiver, one-dimensional linear recording arrays are sufficient to satisfy the gray shade imagery array recording

requirements. If two-dimensional recording arrays were feasible, the image recording rate could be increased by virtue of the "hopping" scanning action appropriate for a two-dimensional recorder. However, the system image throughput rate is limited by the pixel rate of the array sensor rather than by that of the array recorder; in part, this is because the array sensor has to scan and process many source pixels in order to generate one processed pixel ready for recording.

Another major focus of the study effort was the exploitation of the nature of the array scanning technique in order to accomplish sophisticated signal processing functions with low implementation complexity. The nature of the array scanning process lends itself quite naturally to analog sampled-data processing via CCD's. A set of realistic and meaningful systems requirements was adopted in order to define the system and subsystem tradeoffs. It was found that a relatively simple ASITS implementation relying almost entirely upon two types of CCDs manufactured by Reticon corporation is capable of performing all of the desired sophisticated signal processing functions. The ASITS conceptual design presented in Section 7.0 is capable of three BWCR's, three resolutions, image decimation and interpolation by factors of two or four, selection of either image enhancement or MTF correction at full resolution, fixed image enhancement or MTF correction (not selectable) at one-half or one-quarter resolution, and REARCS operation. Computer-assisted evaluation of the effects of CCD time dispersion showed that this property of practical CCD's will not be a limiting factor on system performance. Likewise, computer experiments regarding defective sensor or TAD elements showed that losing an element or row of elements is not catastrophic for the

two-dimensional filter sizes likely to be used in any ASITS implementation. Very extensive computer simulations and subjective evaluation of the resulting imagery showed that two-dimensional nonseparable FIR filters of size  $31 \times 31$  gave very good performance in conjunction with decimation and interpolation. It was found that values of the Kaiser filter window parameter  $\beta$  that give the best performance depend upon the D/I factor. The very considerable reduction in implementation complexity that occurs when separable filters are used provided the impetus for examining the performance of these filters. In all except the most demanding situations, it is felt that separable filters of size  $31 \times 31$  or larger give quite adequate performance. In fact the only region of the source image where a difference between separable and nonseparable filter performance could be discerned was within the region of the frequency wedge corresponding to "diagonal", i.e., off-axis two-dimensional frequencies. The separable filters actually pass more frequencies than the nonseparable filters, and without any additional aliasing degradation. To see why this is so, refer to Figure 8.1 (a), which shows both the two-dimensionally sampled source spectrum (cross hatched) and the sampled data lowpass filter prior to decimating the lowpass filter output. Figure 8.1 (b) shows the spectrum at the filter output after decimation. Note that whether "round" (nonseparable) or "square" (separable) lowpass filters are used, no aliasing occurs. This situation is improved still more by virtue of the fact that spectral energy of images that have been rectilinearly sampled tends to "cling" to the two frequency axes. We verified this fact by plotting a three dimensional perspective of the spectrum of a small portion of the source image.





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Figure 8.1  
Graphical Representation of the Two-Dimensional  
Decimation Process Applied to a Sampled Image Source



STUDY RECOMMENDATIONS

This study has shown that array scanning is both feasible and advantageous in image transmission systems. The application of the array scanning concept to image transmission systems should be further developed to the point of building an engineering model of a full-capability ASITS, with the development proceeding in an orderly fashion over a period of several years. A suggested sequence of future study phases is given below.

- Phase 1: ASITS feasibility study--completed September 1978.
- Phase 2: Detailed end-to-end computer simulation of entire full-capability ASITS resulting in an optimized paper design. Special attention to be given to the comparison between the conceptual design of Section 7.0 and the alternative design that is also mentioned there (uses hybrid technique at all resolutions). Time frame: June 1979 - December 1979.
- Phase 3: Breadboarding of an array sensor with collection of sufficient data to characterize performance. Time frame: January 1980 - June 1980.
- Phase 4: Breadboarding of an array recorder with collection of enough data to characterize performance. Play breadboard sensor to breadboard recorder back-to-back and evaluate performance. Time frame: July 1980 - December 1980.
- Phase 5: Add two-dimensional filtering capability to array sensor implementation and evaluate performance. Time frame: January 1981 - June 1981.

Phase 6: Add hybrid (DCT/DPCM) transformer at transmitter and hybrid inverse transformer at receiver. Also include REARCS capability, which implies various bit assignments for different BWCR's and resolutions. The required along-scan decimation and interpolation filters are the same as those used in Phase 5; this reduces the number of additional R5602's that must be purchased for this phase. The result of this phase is a full-capability ASITS. Extensive evaluation and experimentation with this system will help specify the final design of a production model ASITS. Time Frame: July 1981 - July 1982.

Two or more of the phases suggested above may be combined in order to compress the ASITS development schedule if desired.

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### LIST OF ABBREVIATIONS

A/D = Analog-to-Digital  
AO = Acousto-Optic  
AOBD = Acousto-Optic Beam Deflector  
AOM = Acousto-Optic Modulator  
AOPC = Acousto-Optic Page Composer  
ASITS = Array Scanning Image Transmission System  
BBD = Bucket Brigade Device  
BWC = Bandwidth Compression  
BWCR = Bandwidth Compression Ratio  
CCD = Charge Coupled Device  
CdS = Cadmium Sulfide  
CID = Charge-Injected Device  
CMD = Command  
CTD = Charge Transfer Device  
CTI = Charge Transfer Inefficiency  
D/A = Digital-to-Analog  
DCT = Discrete Cosine Transform  
DEFT = Direct Electronic Fourier Transform  
DFT = Discrete Fourier Transform  
D/I = Decimation/Interpolation  
DIP = Dual In-Line Package  
DPCM = Differential PCM



LIST OF ABBREVIATIONS (CONT'D)

FFT = Fast Fourier Transform  
FIR = Finite Impulse Response  
IIR = Infinite Impulse Response  
IITS = Intra-Threater Image Transmission System  
LED = Light Emitting Diode  
LPF = Low Pass Filter  
lpi = lines per inch  
LSB = Least Significant Bit  
MADD = Multiply-and-Add  
MSB = Most Significant Bit  
MTF = Modulation Transfer Function  
NEP = Noise Equivalent Power  
PCM = Pulse Code Modulation  
PTA = Phototransistor Array  
Pixel = Picture Element  
PR = Pixel Rate  
P/S = parallel-to-serial  
PSF = Point Spread Function  
RAM = Random Access Memory  
REARCS = Redundant Area Coding System  
RF = Radio Frequency  
ROM = Read-Only Memory

LIST OF ABBREVIATIONS (CONT'D)

SAW = Surface Acoustic Wave

S/P = Serial-to-Parallel

SR = Swath Scanning Rate (inches/sec)

1D = One Dimensional

2D = Two Dimensional

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